

Genetics of General Cognitive Ability

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ABBREVIATIONS

AD	Alzheimer's Disease
AIP	Allele image pattern
AMPA	α -amino-3-hydroxy-5-methyl-4-isoxazolepropionate
APOE	Apolipoprotein E
AST	Allele specific test
bp	Basepair
CaMKII- α	Calcium/calmodulin dependent protein kinase II alpha
cM	Centimorgan
DHPLC	Denaturing High-Performance Liquid Chromatography
DNA	Deoxyribonucleic acid
dNTP	Deoxynucleotide
ddNTP	Dideoxynucleotide
DZ	Dizygotic
EEA	Equal environment assumption
EEG	Electroencephalography
GRR	Genotype relative risk
HRR	Haplotype relative risk
IQ	Intelligence quotient
kb	Kilobase
LD	Linkage disequilibrium
LOD	Logarithm of odds
LTD	Long term depression
LTP	Long term potentiation
MRI	Magnetic resonance imaging
MZ	Monozygotic
NCV	Nerve conduction velocity
NMDA	N-methyl-D-aspartate
PCR	Polymerase chain reaction
QTL	Quantitative trait locus
RFLP	Restriction fragment length polymorphism
RNA	Ribonucleic acid
SNP	Single nucleotide polymorphism
STR	Short tandem repeat
UTR	Untranslated region

SUMMARY

General cognitive ability (g) is a general mental ability to reason, solve problems, comprehend complex ideas, think abstractly, learn quickly and learn from experience. Currently used IQ tests are excellent predictors of g . Heritability estimates for g range between 0.30 and 0.75 making it one of the most heritable human behavioural traits. Many behavioural phenotypes, including g , can be described as complex traits. Inheritance of such traits is governed by a mixture of genetic and environmental factors. Genetic factors contributing to total variance in g are likely to be numerous and additive in nature. In order to identify some of the genetic loci contributing to the total variance in g two approaches were employed. First, a genome-wide association study and second, candidate gene study. Genome-wide association study involved testing 1847 microsatellite markers with an average spacing of 2cM. Markers were initially screened on "original" DNA samples. This was followed by testing all positive findings on an independent "replication" sample set. Only one marker, D4S2460, was significant when all the stages of the study were completed. Investigation of candidate genes involved testing of known Apolipoprotein E (ApoE) promoter polymorphisms and Calcium/calmodulin dependant protein kinase II alpha (CaMKII- α) polymorphisms identified in our laboratories through mutation detection techniques. None of the polymorphisms in either of the two genes showed statistically significant association with a general cognitive ability.

Genetics of General Cognitive Ability

1 INTRODUCTION

“Of the high importance of the intellectual faculties there can be no doubt, for man mainly owes to them his predominant position in the world.”⁸

1.1 *What is intelligence ?*

It would be appropriate to begin with a scientific definition of intelligence. However, finding a single definition applicable to all times, places and cultures proved to be a difficult task for several reasons. First, intelligence is unlikely to be a single process with easily observable boundaries, detached from other human faculties such as learning and memory ¹⁸. Second, different cultures have different conceptions of intelligence; designating as “*intelligent*” sets of cognitive, social, and behavioural attributes valued as adaptive to the requirements of living in those cultures ²¹. Third, intelligence tends to be defined differently across different scientific disciplines, and the issue becomes even more complicated by the introduction of theories of multiple intelligences ²².

So, to answer the question “what is intelligence?” and by doing so to lay the foundation for this study, it seemed reasonable to choose a definition regarded as

mainstream among leading researchers on intelligence. The mainstream statement, first published in 1994, says that intelligence is a general mental capability that involves the ability to reason, plan, solve problems, think abstractly, comprehend complex ideas, learn quickly and learn from experience²³.

1.2 History of intelligence research

1.2.1 Early beginnings

Variation in intelligence became the subject of scientific study in the last half of the 19th century. Charles Darwin's theory of evolution by natural selection inspired most of the work carried out in those early days of intelligence research. Darwin introduced the idea of advancement in intellectual capacity through natural selection and transmission of inherited intelligence as a key step in human evolution. His younger cousin, Sir Francis Galton, set out to prove this by collecting data on the intellectual capacities of successful families in Britain at that time. Galton's first impressions on the subject and strong belief that "talent is transmitted by inheritance in a very remarkable degree" were published in 1865²⁴, but most of the data and evidence that intellectual superiority runs in families were presented in "*Hereditary Genius*"²⁵. This book opened a permanent and sometimes heated debate over the relationship between intelligence and heredity.

Galton was fully aware of the difficulties of distinguishing between the effects of heredity (Nature) *versus* those of the environment (Nurture) and chose a powerful tool – twins. His early studies were based on a substantial collection of anecdotal stories about twins²⁶, but recognising the weakness of the approach he decided to devise a

test that would objectively measure differences in mental abilities. This test was meant to probe the sensitivity of perceptions by measuring sight, hearing, sensitivity to slight pressure and response to simple stimuli. Even though such measures of intellectual ability were soon discredited, Galton's achievements in the field continued to influence research on intelligence to this day.

In a similar attempt to assign exactness and consistency to measurements of intellectual abilities, James McKeen Cattell suggested a series of ten mental tests aimed at measuring individual differences in fundamental mental processes ²⁷. There are certain attributes that a person must have to be a successful university student and intelligence is surely one of those. However, failing to predict university students' grades, Cattell's tests were found to be an unreliable tool for measuring intelligence ¹⁸.

1.2.2 20th Century

The beginning of the 20th century marked another turning point in intelligence research mostly owing to Alfred Binet and his contribution to mental test design. He believed that the ability to "judge well, to comprehend well and to reason well" together with the ability of "adapting one's self to circumstances" are essential to natural intelligence and that mental tests should be designed to probe for these rather than sensory acuity ²⁸. The Binet scales, as we know them today, formed the basis of modern IQ tests ²⁹.

Most of the further development of IQ tests was carried out by Henry Goddard, Lewis Terman and Robert Yerkes. In 1916 Terman produced a revised version of Binet's test, the Stanford-Binet test, that became a standard against which all later IQ tests were judged. This test is administered individually by a trained person, which is not only time consuming but has a potential problem of introducing variation in scoring.

Yerkes saw the advantage in having a test that could be administered to a large number of people in a short space of time and introduced the first version of it for the purpose of selecting US Army recruits upon America's entry into the First World War. By the 1920s such IQ tests were widely used in primary and secondary schools as well as by some universities as an entrance requirement.

Intelligence research entered a difficult period during the 1920s when misuses of IQ testing started to occur. Goddard, Terman and Yerkes were firm believers that feeble-mindedness was hereditary and that mating between such individuals should not be allowed:

"Feeble-mindedness is hereditary and transmitted as surely as any other character. We cannot successfully cope with these conditions until we recognise feeble-mindedness and its hereditary nature, recognise it early, and take care of it." ³⁰.

In response, Virginia, followed by a number of other American states, passed a 1924 Eugenic Sterilization Act based on a Model Eugenic Sterilization Law published a decade earlier ³¹. Sterilisation of people in institutions for the mentally ill and retarded continued through the mid 1970s resulting in more than 60,000 Americans enduring involuntarily sterilisation on the grounds of mental retardation ³².

Racial and ethnic differences in IQ scores were yet another sensitive issue. Controversy started after Goddard discovered that a high proportion of immigrants from eastern and southern Europe had IQ scores below average and used this as evidence that differences in intelligence between races are hereditary. This period in American history was difficult, both economically and politically, and politicians readily embraced Goddard's argument to support their own oppressive ideas. In an attempt to reduce the

number of immigrants, particularly those from "black-listed" countries they speeded up the legislation of the Immigration Act of 1924. Critics responded immediately pointing to the cultural bias of the tests that placed immigrants to the United States at a distinct disadvantage.

The Second World War marked another period of advance in test design. Even though it was well known that tests were capturing individual differences in intellectual capacities, no-one was attempting to use testing results to promote discriminatory laws. Attacks on testing faded.

Environmentalism seriously hampered intelligence research during the 1960s. This was the period when leading psychologists fiercely argued against the role of genes in the origins of individual differences in mental abilities. The reaction to Arthur Jensen's article (1969) on genetic influence on ethnic differences in IQ scores was immense. Jensen's monograph led to unjust criticism of all behavioural work carried at that time bringing intelligence research almost to a halt. One beneficial consequence of the criticism was that it induced a series of bigger and better designed studies that used family, adoption and twin designs. The large amount of good quality data from such studies contributed to a final shift in scientific thought that occurred in the 1980s. It brought the acceptance of the idea that differences in intelligence are significantly associated with genetic differences between individuals³³.

1.3 General cognitive ability

1.3.1 Galton's attempt

The prevailing belief during the 19th century was that the mind is a combination of a number of separate functions, called *faculties*. Faculty psychology assumed the existence of a distinct faculty for each and every mental activity¹⁰. Unlike most, Galton firmly believed that differences in general mental ability, rather than in faculty psychology, underlie individual differences in mental performance.

He hypothesised that differences in reaction times to auditory and visual stimuli would provide explanation of individual differences in higher and more complex mental functioning. Galton had measured the reaction times of thousands of subjects in an attempt to provide support for the hypothesis. Unfortunately, the large amount of data failed to show correlation with any criteria of intellectual distinction and his rejection of faculty psychology remained based mainly on general impressions rather than empirical evidence.

Charles Edward Spearman (1863-1945), much impressed by Galton's hypothesis, was puzzled by his inability to provide scientific evidence to support the idea. After re-analysing Galton's data by modern statistical methods he managed to show that the main reason for Galton's failure to prove correlation between simple and complex mental processes was in essence measurement error. Without reliable measures it was impossible to show any substantial correlation.

Probably the greatest of Spearman's contribution to intelligence research was the invention of a powerful statistical tool, factor analysis, which led to the discovery of *g* and publication of his two-factor theory in 1904. Proper explanation of the meaning of *g*, as defined by Spearman, requires some understanding of factor analysis, the principles of which will now be outlined.

1.3.2 Principles of factor analysis

For any given set of variables it is possible to construct a correlation matrix. A *matrix* is an array of numbers arranged in rows and columns. If these numbers are correlations of each variable with every other variable, the matrix is called a *correlation matrix*. When all correlations in the matrix are positive, it is called a *positive manifold*¹⁰. The technique of factor analysis is a procedure aimed to simplify complex correlation matrices with a large number of variables in order to allow us to see if there is any emerging pattern in correlations otherwise hard to detect in raw data¹⁸.

Spearman generated one of the first matrices of this kind from data on students' ranks based on ratings in five subjects: Classics, French, English, Mathematics and Music plus a measurement of pitch discrimination (carried out by Spearman himself). He reasoned that if only one common factor can explain all the correlations among variables, the matrix will show a hierarchical order and the correlation coefficients will decrease in size the further away they are from the upper left corner (Table 1)¹⁰.

Variable	Classics	French	English	Math	Pitch	Music	<i>g</i> loading
Classics		0.83	0.78	0.70	0.66	0.63	<i>0.958</i>
French	0.83		0.67	0.67	0.65	0.57	<i>0.882</i>
English	0.78	0.67		0.64	0.54	0.51	<i>0.803</i>
Math	0.70	0.67	0.64		0.45	0.51	<i>0.750</i>
Pitch	0.66	0.65	0.54	0.45		0.40	<i>0.673</i>
Music	0.63	0.57	0.51	0.51	0.40		<i>0.646</i>
Mean <i>r</i>	<i>0.720</i>	<i>0.678</i>	<i>0.628</i>	<i>0.594</i>	<i>0.540</i>	<i>0.524</i>	

Table No. 1: Spearman's correlation matrix and *g* loadings¹⁰

Spearman proposed a method of *vanishing tetrad differences* as means of testing whether the single factor is responsible for all the correlations observed. A simple equation applied to each of the tetrads within the matrix produces a series of results which should approach zero (from either the positive or negative side) if the variables have only one factor in common (Figure 1) ¹⁰.

tetrad a	tetrad b	tetrad c	tetrad d
0.78 0.67	0.67 0.64	0.83 0.66	0.79 0.63
0.70 0.67	0.57 0.51	0.57 0.40	0.45 0.40
	a) $(0.78 \times 0.67) - (0.70 \times 0.67) = 0.054$		
	b) $(0.67 \times 0.51) - (0.57 \times 0.64) = -0.024$		
	c) $(0.83 \times 0.40) - (0.57 \times 0.66) = -0.044$		
	d) $(0.70 \times 0.40) - (0.45 \times 0.63) = -0.003$		

Figure No. 1: Calculating tetrad differences ¹⁰

The fact that correlation between different variables is not the same means that variables contain the common factor to various degrees, in other words they have different *factor loadings*.

1.3.3 Extraction of Spearman's g

When all the variables used in the correlation matrix can be characterised as mental abilities (as in Table 1), the single factor they have in common Spearman termed the **general factor, general intelligence** or g ³⁴.

Testing how well g can explain the observed correlation between two mental ability test scores is performed by reconstituting their correlation coefficients. This is done by multiplication of tests' g loadings. Subtracting reconstituted correlations from initially observed correlations for the whole matrix gives us a *residual matrix*. The residual matrix shows what is left (correlation wise) after the g factor is removed (Table 2)¹⁰.

Variable	Classics	French	English	Math	Pitch	Music
Classics		-0.01	0.01	-0.02	0.01	0.01
French	-0.01		-0.04	0.01	0.06	0.00
English	0.01	-0.04		0.04	0.00	-0.01
Math	-0.02	0.01	0.04		-0.05	0.03
Pitch	0.01	0.06	0.00	-0.05		-0.03
Music	0.01	0.00	-0.01	0.03	-0.03	
Mean r	<i>0.001</i>	<i>0.002</i>	<i>0.000</i>	<i>0.000</i>	<i>-0.004</i>	<i>-0.002</i>

Table No. 2: Residual correlation matrix after g is removed¹⁰

If no other factor/factors in addition to g can be extracted, as was the case with Spearman's matrix, the residual matrix will have values approaching zero in all fields. Any significant residual correlation would mean that more than one factor is implicated in explaining correlation between variables.

1.3.4 Spearman's two-factor theory

After calculating the size of the g contribution to the total variance in six variables examined, it became obvious to Spearman that only about two thirds of it could be explained by g . The rest he attributed to *specific factors*.

Spearman's famous two-factor theory states that individual differences on any measurement of any mental ability are due to only two factors uncorrelated with each other: a general factor, g , common to all mental abilities and a specific factor, s , specific to each and every mental ability measurement³⁴. So, tests for specific mental abilities do measure those abilities, but they all reflect g to varying degrees as well. Years later, in an attempt to explain the meaning of g and s , Spearman compared g to the power generator and s to the different engines, each of which is powered by g but also has its own level of efficiency independent of g ³⁵.

How well the two variables (*i.e.* scores on two different mental ability tests) will correlate is highly conditional on the level of their g loadings. The more complex the cognitive process measured is, the greater the correlation between the test and g and its contribution to overall g score³⁵. Therefore, if one wants to assess a person's level of g it is more efficient to choose a highly g loaded test.

Spearman invested a substantial amount of work in developing a test that could potentially be administered to any individual older than three, regardless of cultural and educational background. The test was further developed by his student, John Raven,

and is now known as Raven's Progressive Matrices. Of all the tests measuring mental abilities in man, this test has the highest g loading (around 0.80) and is often used as a marker test for Spearman's g ¹⁰.

1.3.5 g and s and something else

It was not long after the publication of Spearman's "two factor theory" that its critics started to emerge. Sir Cyril Burt (1883-1971) was one of the leading psychologists of the time who argued that a general factor and specific factors cannot explain all the correlations between various mental ability tests. He suggested that in addition to two extremes, g and s , there are factors that are common to some but not all variables³⁶. Those were later termed *group factors*.

In support to his argument Burt offered a substantial amount of data collected on a number of different tests and showed that the existence of group factors was the rule rather than the exception. Other psychologists agreed that Burt's introduction of group factors in addition to specific factors and g was more acceptable than Spearman's quite rigid two-factor theory. This marked the birth of the first multiple factor theory and the start of new era in intelligence research.

1.3.6 Models and characteristics of g

When performed on correlation matrices composed of a variety of tests of cognitive ability, factor analysis may give different explanations of the relationship between g and

the other factors. These different ways of looking at those relationships are termed "models of g "¹⁰. There are four basic models of g :

(a) The Spearman Model

It has the simplest structure of the four. The general factor, g , is present in all variables and is the only factor they have in common. If the method of vanishing tetrads shows that more than one factor can be extracted, the model will give an incorrect value for g . See Figure No. 2.

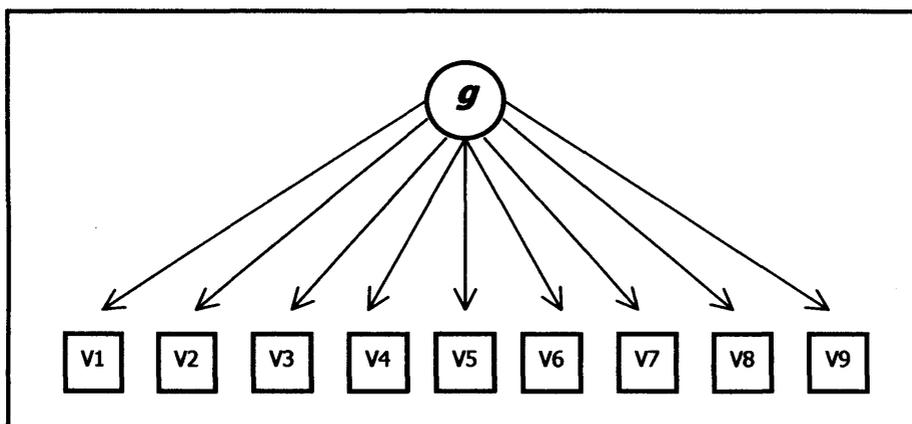


Figure No. 2: The Spearman model of g

(b) The Thurstone Model

This multifactor model proposes existence of group factors, but no general factor common to all the variables. The model never held up even when applied to mental tests, devised by Thurstone himself, aimed to measure primary mental abilities as purely as possible. All the variables correlated with each other because they all measured g in addition to primary mental abilities¹⁰.

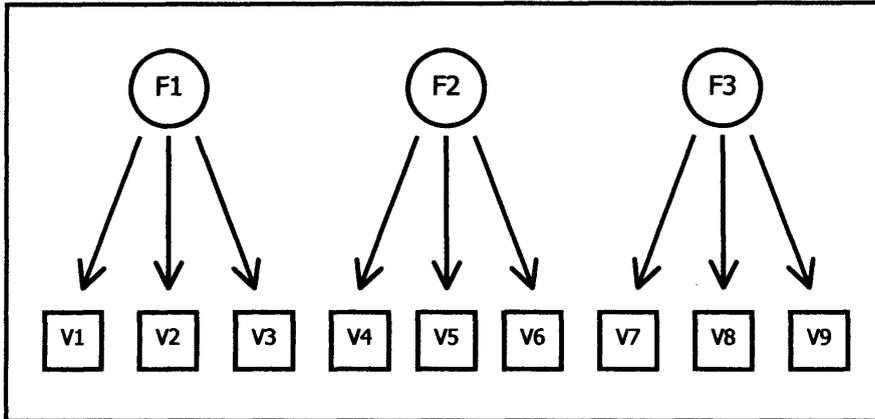


Figure No. 3: The Thurstone Model of g

(c) The Bifactor model

The model is characterised by existence of both Spearman's g and Thurstone's group factors. Variables' correlations with group factors and g are independent from each other.

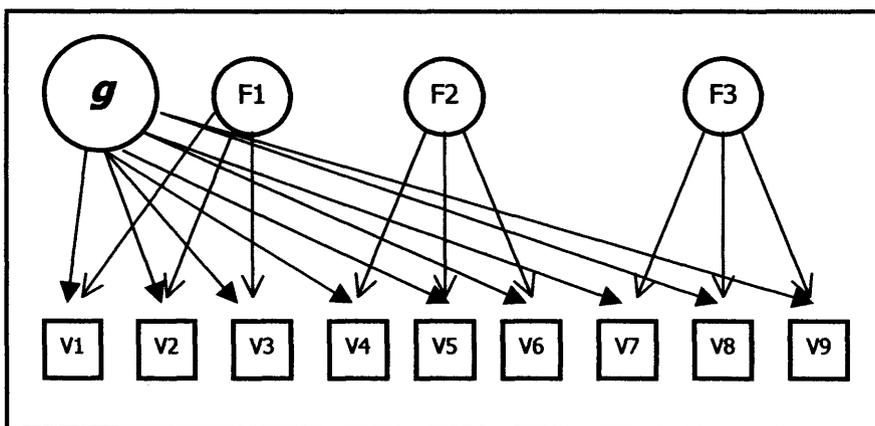


Figure No. 4: The Bifactor model of g

(d) The Orthogonalised Hierarchical Model

This is the preferred model of g . It consists of variables, at the bottom, first-order factors above and a second-order factor, g , at the top (Figure No. 5). Correlation between variables gives rise to first-order factors (group factors), while correlation between group factors (F) produces g . Typically, g accounts for a larger proportion of the total variance than all of the other factors combined. Each of the variables is also characterised by uniqueness (u), which is the variance not explained by any of the higher-order factors ¹⁰.

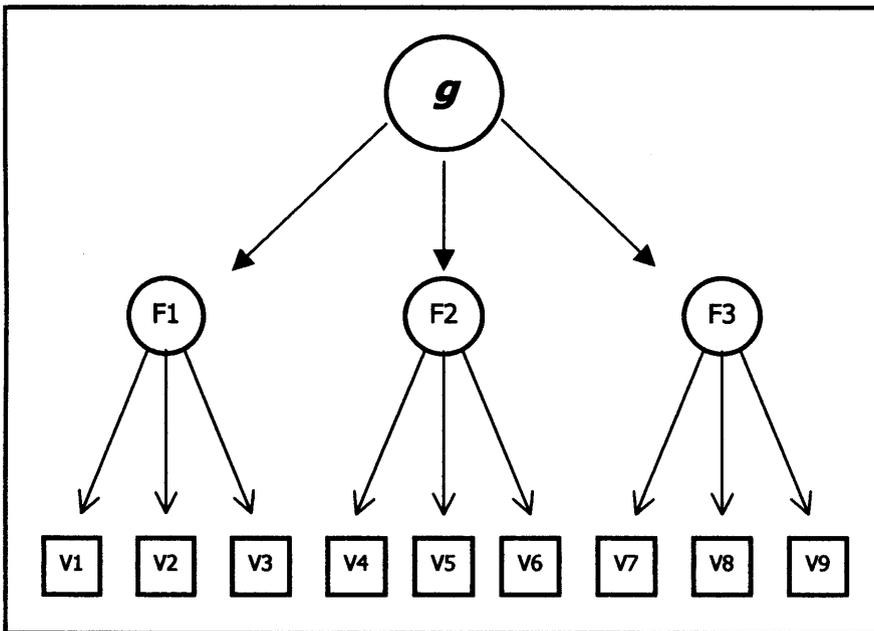


Figure No. 5: The Orthogonalised hierarchical model of g

1.3.7 Consistency of g

The basis of g , as described above, is that the correlations between a number of cognitive ability tests are all positive. How invariant g will be depends highly on the collection of tests used and the method of factor analysis applied to the correlation matrix. In other words, no method of factor analysis will produce exactly the same value of g when applied to different test batteries, and there are no two methods of factor analysis that will give the same value of g even when applied to the same set of mental tests ¹⁰.

This can be explained by the fact that when creating a set of tests we are not able to achieve a perfect representation of all possible cognitive tests. Instead, we will have a limited sample of tests that will yield somewhat different g than another limited sample. It has to be stressed, though, that these variations in g are very small and that a very similar g emerges from most collections of quite diverse cognitive tests. This observation is of great importance, because it gives a substantial weight to the test constructors' claims that their tests do measure something (general intelligence) that is essential for good performance on a variety of mental tests ¹⁸.

Another interesting question is whether the g loadings of tests are the same for subjects who differ in general intelligence. Detterman and Daniel (1989) reported a study that showed how average correlation among the tests declines as the general ability level increases ³⁷. Their data indicated that the amount of variance in intelligence attributable to g varies inversely with general intelligence ³⁷.

1.3.8 Relationship between g and IQ

The intelligence quotient (IQ) refers to a score on a mental ability test. Characteristically, raw scores obtained are standardised within narrow age intervals on a large random sample of a specified population. This produces IQ scores that will have same mean (100 points), standard deviation (15 points) at every age level, and approximately normal distribution in the standardised population. Approximately 95% of the population has scores within two standard deviations of the mean, between 70 and 130 (Figure No. 6) ³⁸.

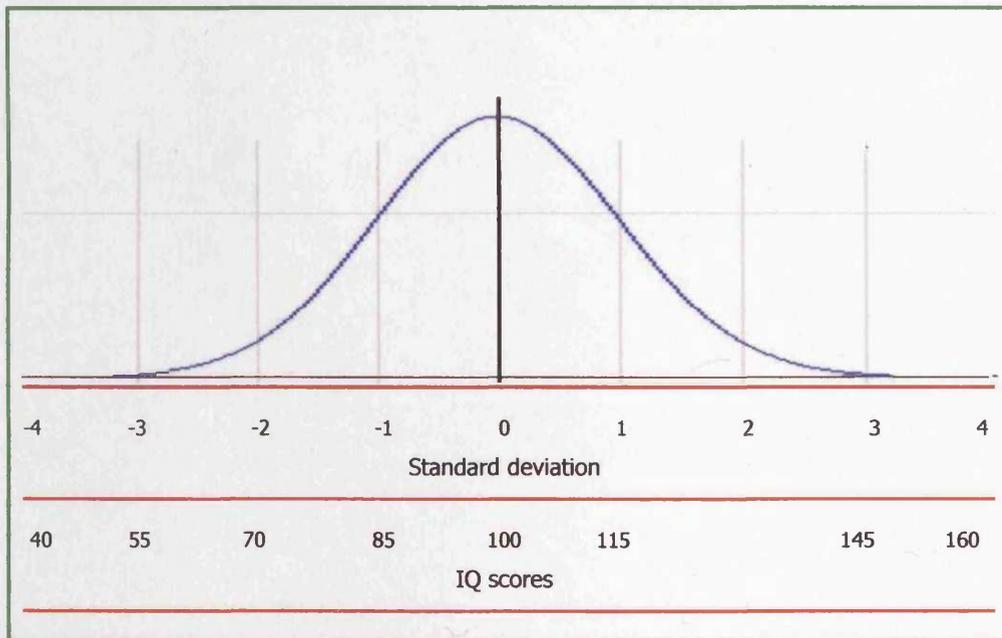


Figure No. 6: Population distribution of IQ scores

The main characteristic of all IQ tests is that they are all highly *g* loaded. The most *g* - rich items in the test will be those that involve some sort of deductive reasoning, spatial visualisation, quantitative reasoning and verbal reasoning ¹⁰. The best predictors of *g* will be tests with the highest *g* loading (*i.e.* Raven's Progressive Matrices) and very small loadings on less general factors (*i.e.* group factors) ¹⁸.

1.3.9 *g* and educational attainment

Intelligence tests do predict school performance fairly well. Correlations between IQ scores and academic performance usually range around 0.50 ³⁹. A few studies have investigated the relationship between intelligence and academic achievement ^{40, 41, 42, 43}. They provided strong evidence for the hypothesis that all tests of intelligence measure intellectual achievements that are determined by the ability to acquire knowledge and are therefore tests of achievement ³⁹. The nature of correlation between intelligence and educational attainment is such that IQ scores at one age predict educational success at a later age, as shown by independent longitudinal studies performed in Britain ⁴⁴ and America ^{45, 41}.

So, it seems reasonable to say that mental ability tests measure something that is related to success in one aspect of our lives where intelligence has traditionally been thought to be relevant ¹⁸.

1.3.10 The biology of cognitive ability

An important approach to studying intelligence is to try to understand the biology behind it. Intellectual activities surely involve brain functioning and differences in such activities are likely to be reflected in differences observable in the brain. It is difficult to say which one of the number of biological theories best accounts for individual differences in intellectual performance. Quite possibly, a true biological explanation of intelligence will be a combination of several theories.

1.3.10.1 Early biological theories

The first biological theories started to emerge during the 1950s. Hebb, the most influential psychologist of that time, proposed the existence of two types of intelligence: A, a person's innate potential and B, a product of actual brain development ²¹. He also suggested that learning, which is an important basis of intelligence, is achieved through cell assemblies and through creating more and more complex connections between neurons ²¹.

1.3.10.2 Brain activity

More recent theories deal with more specific aspects of neuronal functioning. Two main approaches dominate the field: (1) correlation between electroencephalography (EEG) traces and intelligence, and (2) speed of processing.

The first method examines the relationship between intelligence and brain activity as recorded by electrodes positioned on the head. A common finding in such studies is a negative association between brain activity under cognitive load and intelligence ⁴⁶. In other words, intellectually competent individuals when confronted with complex tasks are more likely to exclude irrelevant neural networks, thereby showing lower levels of brain activity than intellectually less competent individuals. Another interesting observation is that individuals with higher IQs display higher brain activity in resting conditions which is usually explained by the ability of such individuals to engage in complex mental activities when resting and/or dealing with less demanding tasks ⁴⁶.

1.3.10.3 Speed of processing

The second method is based on a suggestion that high cognitive ability is associated with faster information processing as a result of more rapid and efficient central nervous functioning ⁴⁷. Speed of processing is usually estimated through measurements of either peripheral nerve conduction velocity (NCV) or brain nerve conduction velocity (reaction time – RT).

Several studies have found peripheral NCV, the speed at which an impulse travels along the nerve, to be positively correlated with general intelligence (0.20 ⁴⁸ and 0.46 ⁴⁹). Interpretation of these findings requires much caution since there have been several failures to replicate ^{50, 51}.

Reaction time (RT) measures the speed at which subjects respond to visual stimuli by pressing a button on a response console. RT has been found to be moderately negatively correlated with IQ scores (-0.43 to -0.66), that is higher IQ scoring individuals have shorter RTs ^{49, 51}. Unlike the correlation between NCV and IQ, RT is

more consistently correlated with IQ and this has been regarded as a more important factor in explaining individual differences in intellectual ability ⁵⁰.

1.3.10.4 Glucose metabolism

Several studies have investigated the relationship between cortical glucose metabolic rates and mental activities of individuals with different IQ scores. Haier (1988) showed that glucose metabolic rates during performance in complex mental tasks were lower in individuals with higher IQ scores ⁵². This suggested that more intelligent people have to invest less effort than less intelligent ones to solve reasoning problems.

Interpretation of these and similar findings may not be straightforward since it would be difficult, at this moment, to distinguish whether the difference in metabolic rate is a cause or an effect, or whether both high IQ and low glucose utilisation are related to a third variable ²¹.

1.3.10.5 Brain size and structure

Several major studies using magnetic resonance imaging have shown a positive correlation (0.40 is taken to be the best estimate to date) between *in vivo* brain size and intelligence ^{53, 54}. Despite some failures to replicate in studies with monozygotic twins ⁵⁵ this correlation is generally thought to be genuine ⁵⁶. Moreover, an interesting sex difference has been observed in MRI brain studies: female brain size correlates more highly with verbal ability, while correlation figures in males are higher for spatial ability ⁵⁰.

What is it about the size of the brain that is related to an individual's mental ability?

Miller (1994) suggested that larger brains in more intelligent individuals can be explained by greater thickness of the myelin layer surrounding each nerve cell ⁵⁶. Myelination is known to occur during early childhood which is consistent with the improvement in intellectual functioning that occurs as infants mature. Also, less myelin can lead to more cross-talk along neurons, which could be a source of errors in signal transduction leading to lower intelligence ⁵⁷. Myelin degradation is associated with the normal ageing processes and is a possible explanation for general slowing in cognitive abilities with age ⁵⁶.

Another structural feature that is genetically regulated and implicated in intellectual functioning is the distribution of grey matter across the cortex. Thompson *et al.* (2001) examined 20 MZ and 20 DZ twin pairs and found that the quantity of frontal grey matter is more similar in individuals who are genetically alike ($r=0.9$, $p<0.0001$ after correction for multiple testing) and that individual differences in brain structure are tightly correlated with individual differences in IQ scores ($p<0.0176$ after correction for multiple testing) ⁵⁸.

1.3.11 Genetic models of cognitive abilities

Nearly all genetic research on intelligence is based on the so-called psychometric model ³³. This holds that cognitive abilities are positioned at three levels that are related to each other and organised in an hierarchical fashion (Figure No. 7). The psychometric model is based upon, and provides an explanation for, phenotypic relationships between various mental abilities. In terms of genetics, three different

versions of the model have been proposed: 1) *the bottom - up*, 2) *the top - down* and 3) *the level-of-processing* model³³.

The first model, bottom-up, assumes that there are genes that influence elementary abilities, the effects of which then feed into specific cognitive abilities that finally converge on general cognitive ability. The top-down model assumes that genes that influence specific and elementary abilities do so by affecting some general cognitive process like *g*. Finally, the level-of-processing model proposes that different genes are involved at different levels of processing within the hierarchy, but it also allows for the possibility that there are genes common to all cognitive levels within the pyramid.

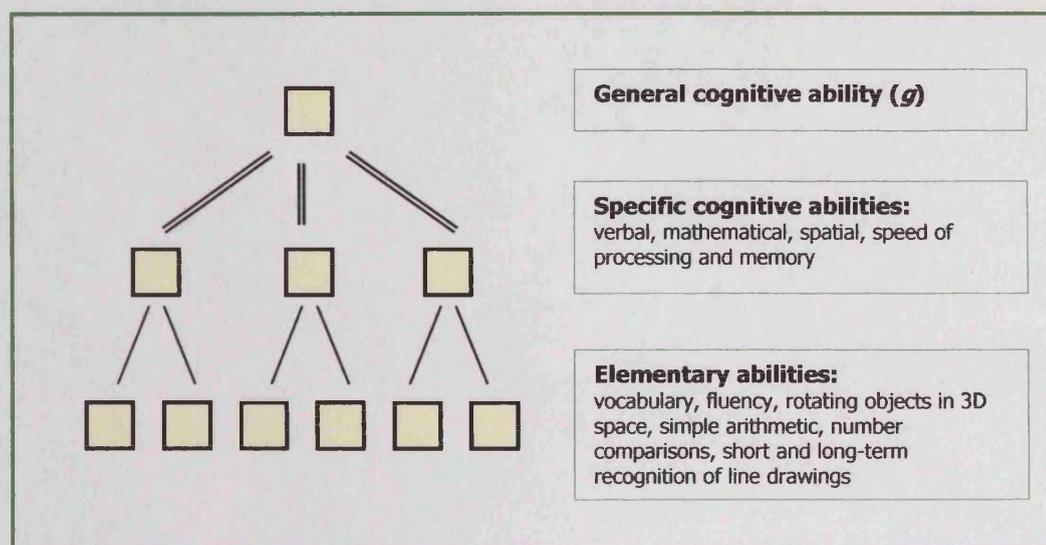


Figure No. 7: Hierarchical organisation of cognitive abilities

Multivariate genetic analyses in the field of intelligence research mainly focus on the relationship between general and specific cognitive abilities, searching for the source of their covariance. So far, the results imply that, to a considerable degree, genetic effects on cognitive abilities are general, whereby genes that affect one cognitive ability also

affect other cognitive abilities (top–down model). There is also evidence that some genetic effects are unique to each specific cognitive ability, as assumed by the level-of-processing model^{59, 60}. The results of most comprehensive analyses incorporating all three levels of processing (general, specific and elementary) lean more towards the level-of-processing model⁶¹.

Genetic multivariate analysis of available data thus predicts that when genes influencing *g* are identified, most of these genes will be associated with specific and elementary abilities as well, while some will be specific to certain cognitive abilities.

1.4 *General cognitive ability as a genetically complex trait*

1.4.1 Complex traits

Many behavioural and neurological phenotypes of interest, including general cognitive ability, can be described as complex traits. The term “complex trait” refers to any phenotype that does not exhibit classic Mendelian inheritance, either dominant or recessive, that can be attributable to a single genetic locus⁶². A defining feature of complex phenotypes is that no single locus contains alleles that are either necessary or sufficient to produce a particular phenotype. Inheritance of such traits is governed by a mixture of genetic and environmental factors (Figure No. 8).

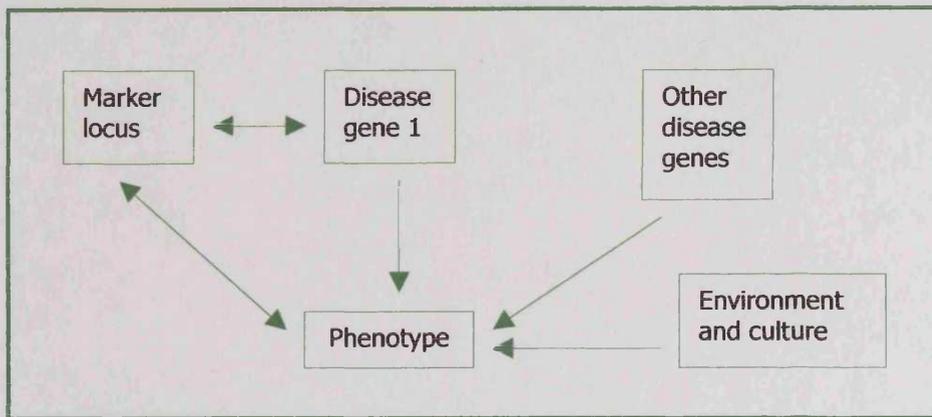


Figure No. 8: A simplified model of the etiological factors predisposing to complex phenotypes ⁴.

To some extent, even a simple genetic disease can be described as complex when closely examined ⁶³. However, such rare Mendelian forms are usually considered separately from more common "complex" forms that aggregate in families, but occur as a result of a combination of multiple genetic effects and environmental factors (multifactorial inheritance).

Different genetic mechanisms can lead to complex inheritance. Broadly speaking, effects on a trait can either be additive or non-additive. Additive genetic effects occur when alleles at different loci add up to affect behaviour, although the size of each allele's contribution can be highly variable ^{64, 63}.

Alternatively, the effect of an allele can depend upon the presence of other alleles and when such interaction between alleles occurs it is called *epistasis*. *Locus heterogeneity* (mutations in any one of several genes resulting in an identical phenotype), *allelic heterogeneity* (multiple alleles existing at a single locus, different combinations of which give rise to different phenotypes), *parent of origin effects* (genomic imprinting) and *incomplete penetrance* (reduced probability that a particular genotype will result in specific phenotype) are additional mechanisms that can

contribute to the complexity of a trait. Finally, additional modes of genetic inheritance such as *genetic anticipation*, a phenomenon of increasing severity and/or decreasing age of onset in successive generations, and *mitochondrial inheritance* can further complicate determination of a precise mechanism operating in any particular trait^{62, 65}.

1.4.2 Quantitative traits

Traits that require the simultaneous presence of alleles in multiple genes are also known as polygenic. They can be further classified as *discrete* or *categorical*, with a specific outcome (*i.e.* development / no development of type I diabetes or asthma), or *quantitative*, measured by a continuous variable (*i.e.* blood pressure, height, weight, IQ scores)⁶².

Each locus contributing to genetic variance in a quantitative trait, termed a quantitative trait locus (QTL), is neither necessary nor sufficient for development of a trait⁶⁶. QTLs denote multiple genes of varying effect size so that, broadly speaking, a trait can be influenced by a few oligogenes with a moderate effect on the phenotype, or by many polygenes each with very small effect, or by a combination of the two³³.

It would be a mistake to assume that all the loci contributing to a quantitative trait such as general cognitive ability, no matter how many or few of them, are of equal effect size⁶⁴. The magnitude of an effect exerted by any one locus can vary from major, to modest, to small. Various statistical models have shown that the most likely scenario usually involves few modest effect loci explaining much of the phenotypic variance, while the rest of it is attributed to larger number of small effect loci plus environment^{63, 67, 68}.

Obtaining accurate estimates for the overall number of genes that affect a given character is difficult for several reasons. First, only those genes with a sufficiently large

phenotypic effect will be detected. Second, the probability of detecting QTLs is a direct function of the distance between a QTL and the nearest genetic marker. Third, locating QTLs is highly dependent on the size of the population analysed: the larger the population, the greater the likelihood of detecting loci of lesser effect. Fourth, it is likely to be more difficult to detect QTL for traits with low heritability. Finally, using stringent criteria in declaring a significant effect lowers the chances of reporting spurious findings, but at the same time reduces the possibility of detecting small effect size QTLs. All these factors account, to a certain extent, for the underestimation of the total number of contributing loci.

1.4.3 The phenotypic distribution of quantitative traits

In categorical traits, it is thought that a combination of multiple genetic loci only produces a character when a threshold of susceptibility is reached ⁶⁹. For a continuous trait, the more genetic loci contribute to overall variance in phenotype, the more closely will the phenotypic distribution resemble the typical Gaussian (normal) distribution curve. In general, if there are n loci then there will be $2n + 1$ possible phenotypic values (Figure No. 9) ⁹. General cognitive ability with its bell-shaped distribution of IQ scores is therefore likely to be a character influenced by a larger number of genes in combination with environmental factors.

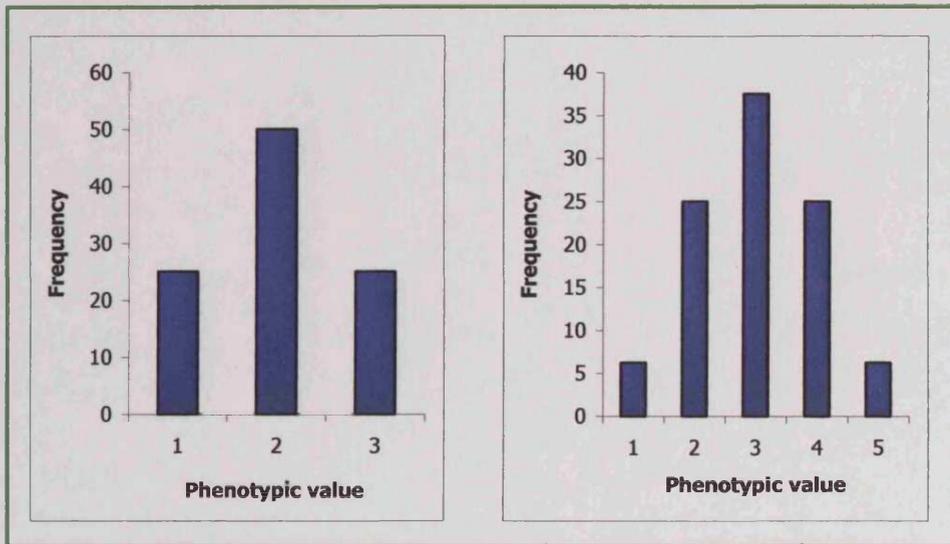


Figure No. 9: As the number of QTLs contributing to a trait increases, the more closely its distribution in the population approximates to a normal distribution ⁹.

As the number of loci increases, the simple relationship between genotype and phenotype disappears ⁷⁰. This is due not only to the larger number of possible genotypes, but also to the fact that alleles underlying quantitative traits are rarely either dominant or recessive. Instead, what is observed is the whole range of gene effects from complete dominance to complete recessiveness, and the dominance/additivity statistic (d/a) is used to describe it. A situation where d/a equals 0 means that a heterozygote is intermediate between two homozygotes and that the genetic contribution to the phenotype is entirely additive ⁷¹.

1.5 Mapping genes for complex traits

1.5.1 Demonstrating a genetic component to a trait

Genetic epidemiological studies are required to provide evidence for the involvement of genes in complex traits. The first clue to the operation of genes often comes from evidence of familial aggregation. However, family studies do not allow the effects of shared environment to be distinguished from those of genetic factors. This requires twin and adoption studies.

1.5.1.1 Family studies

Family studies examine the degree of familial clustering of a trait which can be expressed as λ_r , the ratio of the risk for relatives of patients to the population prevalence of trait (Figure No. 10)⁷².

$$\lambda_r = \frac{\text{Prevalence of the disease in a relative "r" of an affected person}}{\text{Population prevalence of the disease}}$$

Figure No. 10: Calculating relative risk ratio

The risk ratio λ_r decreases with the degree of relationship between proband and relatives⁷³. Traits with an environmental component will show a steeper drop in λ_r with decreasing relatedness than that predicted for a purely genetic character.

Risch (1990) suggested the use of risk ratio as a direct indicator of the mode of inheritance underlying the trait. For monogenic traits λ_r decreases by a factor of two with each degree of relationship⁷³. The same is true for traits where several gene loci are involved in its aetiology as long as they do not interact epistatically. More rapid drops in λ_r are suggestive of multiple genetic loci and epistasis⁷³.

1.5.1.2 Twin studies

Twin studies attempt to distinguish genetic from environmental sources of familial resemblance by comparing the phenotypic resemblance of monozygotic (MZ) twins who share 100% of their genes to the resemblance of dizygotic (DZ) twins who are on average only 50% similar genetically. In addition, both MZ and DZ twins share the same uterus, age and aspects of their early and later environment. This allows the variation of traits to be separated into genetic and environmental components.

If a trait is influenced by several genes with purely additive effect, and if there is no environmental involvement, then the trait should be twice as correlated between MZ twins than between DZ twins³³. This relationship will remain the same if shared environment also plays a role, as long as both MZ and DZ twins experience the same level of shared environmental component⁷⁴.

The twin method relies upon the *equal environment assumption* (EEA) which assumes that environmental similarity is roughly the same for both types of twins reared in the same family³³. Deviation from this simple relationship should be expected

for traits with an environmental component as well as those reflecting more complex gene interactions⁶⁸.

Critics of the method argue that the equal environment assumption is wrong and that, apart from being genetically more similar than DZ twins, MZ twins are often environmentally more similar^{75, 74}. This, in their view, makes classical twin studies of doubtful value as indicators of genetic influence on psychiatric disorders and behavioural traits^{75, 76}.

In an attempt to prove the sceptics wrong, a number of studies have been conducted with the aim of testing the equal environment assumption. One of the earliest such studies examined the effect of physical resemblance of identical and fraternal twins (206 twin pairs) on the ratings for four personality traits⁷⁷. The data suggested that greater resemblance in appearance in identical twins does not make them more similar in personality. Thus, although greater similarity in appearance may result in unequal environments for the two types of twin pairs, it does not appear to bias twin studies in the direction of inflated heritabilities⁷⁷.

Parents may treat MZ twins more similarly than DZ twins, thereby violating the EEA. To test the validity of the hypothesis Kendler *et al.* (1994) conducted a study in which they examined whether parents' beliefs about their twins' zygosity influenced twin resemblance for common psychiatric disorders. In about 20% of cases parents' beliefs about their twins' zygosity, *perceived zygosity*, disagreed with *true zygosity* of the twins⁷⁸. The authors found that even these 'mis-labelled' twins were as similar behaviourally as correctly labelled ones, and that differential treatment of MZ and DZ twins by their parents is unlikely to cause violation of EEA⁷⁸.

Further evidence in support to the traditional twin method was offered by Kendler *et al.* (1993) and Hettema *et al.* (1995) who investigated the impact of physical similarity on phenotypic resemblance in five common psychiatric disorders: major depression, generalised anxiety disorder, phobia, bulimia, and alcoholism^{79, 80}. They examined 1030 and 882 female-female twin pairs respectively, and concluded that the level of

physical resemblance did not have significant effect on concordance for any of the five disorders, thereby providing support for the validity of equal environment assumption ^{79, 80}.

Similar treatment of MZ twins on the basis of their zygosity alone is, therefore, not a threat to the validity of the equal environment assumption as applied to twin studies of psychiatric disorders and behavioural traits ^{81, 82}.

1.5.1.3 Adoption studies

Adoption studies are another way of disentangling genetic from environmental causes of phenotypic variation. Adoption creates pairs of genetically related individuals who do not share a common family environment and pairs of unrelated individuals who do share the same environment. In the absence of selective placement resemblance between adopted children and their adopted parents directly assesses the environmental contribution to the parent-offspring resemblance ³³.

It is true that significant selective placement has taken place in some adoption projects where correlation in IQ scores between adopted children and their adopted parents was 0.34 even before children started to live with their adopted families ⁸³. One way of measuring selective placement is by looking at correlation between adoptive parents and biological mothers, or between biological children of adoptive parents and biological mothers of adopted children. Any significant correlation could not be explained by shared genes, but rather the similar circumstances between biological and adoptive families ¹⁸. If, however, correlation between biological mother and her adopted-away child is significantly higher than correlation between her and biological children of adoptive family, the only reasonable conclusion will be that resemblance

between biological mother and her child cannot be entirely explained by selective placement ¹⁸.

There are three main designs of this type of a study, each addressing a different question ¹⁸.

- a) **Adoptive families vs. biological families:** this design is used for comparing the magnitude of correlation between children and their parents in adoptive *versus* biological families. If correlations are significantly higher between members of biological families, it suggests that they share genetic composition as well as environment.
- b) **Correlations in adoptive families:** correlations in adoptive families significantly greater than zero are suggestive of common family environment being an important contributor to similarities in phenotype examined.
- c) **Adopted children vs. biological parents:** this type of design should provide an answer to question of whether children resemble their biological parents more than adopted parents? A positive answer suggests genetic influence on a trait.

In practice, it is often possible for a single study to generate enough data necessary to provide answers to all of these questions at once ^{84, 85, 86}.

The main obstacles in such studies are the lack of information on biological parents and potential selective placement of children into adopted families. Nevertheless, adoption studies are still the most powerful method for checking whether a character is genetically determined ⁷⁰.

1.5.2 The heritability of g

1.5.2.1 Defining heritability

Quantitative traits, such as g , tend to be normally distributed in the general population. The phenotypic variance (V_p) observed is due to combined action of genetic variance (V_g) and environmental variance (V_e). Both, V_g and V_e can be broken further down. V_g is composed of additive and non-additive genetic effects, while V_e is a product of shared and non-shared (special) environmental effects (Figure No. 11).

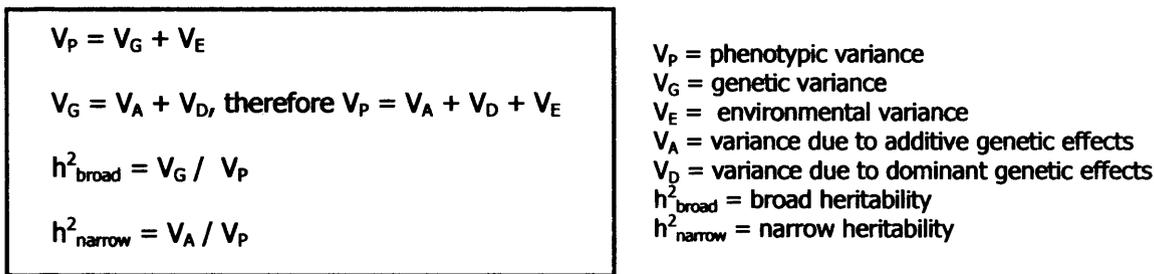


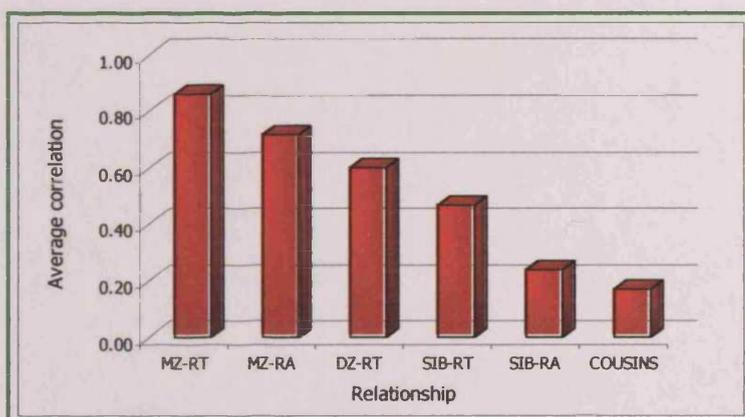
Figure No. 11: Variance components and heritability

The heritability (h^2) of a trait is the proportion of total population variation in a given characteristic that can be explained by genes⁷⁰. It is not immutable, but can vary over time and between different populations and environments. The heritability of a trait depends greatly on the environmental variance. The lower the V_e , the more of the total variation in phenotype can be explained by genetic variation and the more heritable the trait will be.

1.5.2.2 Evidence for a genetic contribution to variance in *g*

Darwinian theory implies that members of one population vary in many characteristics and that much of that variation is inherited. Galton's work on twins was the first attempt to show that this general truth can be applied to human intelligence as well²⁶. However, it took more than hundred years for a sizeable amount of good quality data to be generated.

In order to estimate the heritability of a trait one can look at the degree to which variation in environmental or genetic relatedness influences phenotypic resemblance. Bouchard and McGue published an influential review of genetic epidemiological studies of *g* up until 1981⁸⁷. They compiled correlation data from 111 carefully selected studies clearly showing that the higher the proportion of genes two family members have in common, the higher the average correlation between their IQ scores (average correlation for MZ twins reared together was 0.86, for MZ twins reared apart 0.72, for DZ twins reared together 0.60; a larger list of reported correlations is presented in Figure No. 12). The evidence marshalled and reviewed in this paper has often been used to support the idea that there is a substantial genetic contribution to overall variance in *g*.



MZ-RT: monozygotic twins reared together

MZ-RA: monozygotic twins reared apart

DZ-RT: dizygotic twins reared together

SIB-RT: siblings reared together

COUSINS

Figure No. 12: Dependence of average correlation in IQ scores on genetic relatedness between individuals.

1.5.2.3 Estimating the heritability of g

If MZ and DZ correlations are the same, heritability is estimated to be zero. A correlation between MZ twins of 1.0 and between DZ twins of 0.5, suggests that genetic differences among individuals completely explain their phenotypic differences. Heritability can be estimated approximately from studies involving both MZ and DZ twins. Because MZ twins share 100% of their genes and DZ twins on average only 50%, the difference in their correlations reflects half of the genetic effect. Doubling the correlation difference, the *Falconer heritability formula*, produces a heritability estimate for the trait (Figure No. 13) ². The equation is based on the equal environment assumption and on the assumption that all genetic influence is additive in nature ².

$$h^2 = 2 (r_{MZT} - r_{DZT})$$

Figure No. 13: Falconer formula for calculating heritability ².

Estimating heritability from correlation data requires quantitative modelling and analysis ¹⁸. A good model is based on several assumptions that make it more realistic. In 1990 Chipuer *et al.* provided such a model ⁸⁸. The model acknowledges a distinction between additive and non-additive genetic effects, it assumes that twins share more similar experiences than other siblings and acknowledges the possible importance of assortative mating ¹⁸.

Model-fitting analysis of the correlations reported by Bouchard and McGue ⁸⁷ produced heritability estimates for g of about 50 percent ⁸⁸. Even though most theorists

agree with the 0.50 heritability estimate ^{84, 88, 89}, there are some who prefer a higher figure of around 0.70 ⁹⁰.

Studies involving MZ twins adopted away are important in respect that correlations observed between such individuals provide direct estimates of heritability levels for a particular behaviour dimension ³³. The Minnesota study of twins reared apart examined 45 pairs of monozygotic (MZ) twins adopted away and reported correlation between their IQ scores of 0.78 ⁹¹. Figures between 0.60 and 0.80 for heritability of general intelligence were also reported by others ^{7, 92, 93}. Teasdale and Owen (1984) reported the highest IQ heritability of 0.96 ⁹⁴ using siblings reared together and apart.

Pedersen *et al.* argued that higher broad heritability estimates for IQ are due to epistasis or dominance, and that a model with all non-additive genetic variance fits their data best ⁷. However, other major studies on intelligence did not find evidence of non-additive genetic variance and concluded that such model is an unlikely model for general cognitive ability ^{94, 84, 91, 88}.

Taken together, it is reasonable to say that the broad heritability of IQ in modern industrialised societies is somewhere between 0.30 and 0.75, since currently available data and quantification models do not justify much greater precision ¹⁸.

1.5.2.4 Developmental changes in heritability of *g*

Most people mistakenly believe that the heritability of *g* does not change from the moment of conception. Longitudinal studies are designed to follow individuals for a longer period of time making it possible to analyse changes in the genetic contribution to phenotypic variance. The results of several such studies ^{95, 84, 91, 7} show that the heritability of general cognitive ability increases during the life-span, while the importance of shared environment diminishes in favour of non-shared experiences.

The Colorado Adoption Project was the first large-scale longitudinal study of behavioural development. 245 children were tested at 1, 2, 3, 4 and 7 years of age. Analysis of data provided evidence for increasing heritability of general mental ability from 9% at 1 year of age to 36% at 7 years⁹⁵.

Another change in heritability levels appears to occur at the time when most individuals leave home and start to function as adults on their own, as observed in studies following older subjects³. Finkel *et al.* (1995) combined data from the Minnesota⁹¹ and Swedish studies⁷, and showed that for the age periods 27–50 and 50–65 estimated IQ heritability is approaching 0.80 (Fig. No. 14)⁹⁶.

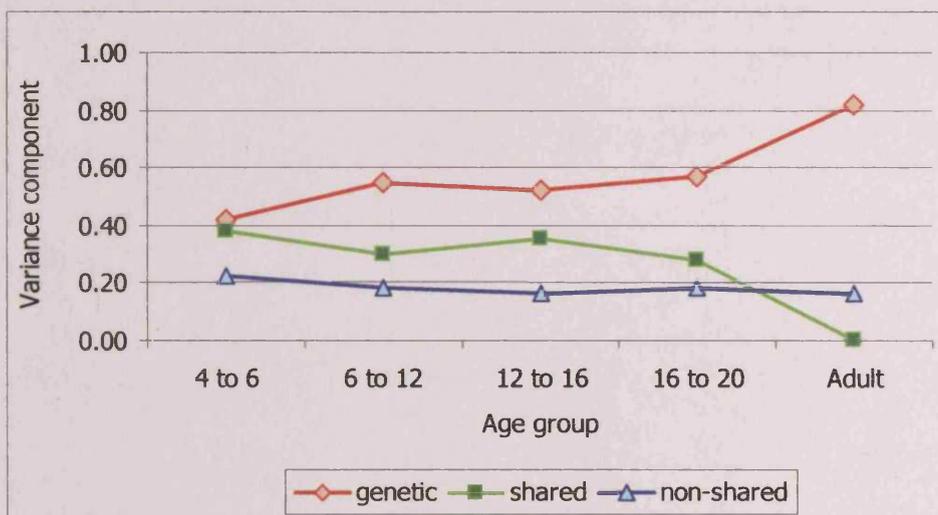


Figure No. 14: Effect of age on genetic and environmental sources of variation in g^3 .

Yet, inferences have to be made with great caution. Genetic relatedness can cause similarity in experiences, which feeds back to produce similar IQ scores; so-called *gene-environment correlation*³³. There are three types of genotype-environment correlation: passive, reactive (evocative) and active⁹⁷.

Passive: Children inherit environments from their parents that are correlated with their genotypes; for example, if musical talent is heritable, then musically gifted children are likely to have musically gifted parents that will provide environment necessary for development of musical ability.

Reactive: Individuals provoke reactions from other people on the basis of their genotypes; for example, musically talented children may be recognised at schools and given special opportunities for further development of the ability.

Active: Individuals seek or create environments that are correlated with their genotypes; for example, musically gifted children may seek out musical environments.

Since such effects might be cumulative, this might account for the increased correlation between IQ scores among twins with time ¹⁸. So, children showing high correlation in IQ scores with their adoptive parents early in childhood will start creating their own environments and experiences as they grow older, driven by their genetic backgrounds. This will cause their and their adoptive parents' and siblings' IQ scores to start diverging and go further apart as they are approaching adulthood ⁸⁴.

Gene–environment correlation is not clear cut. Therefore, estimating the heritability of g is bound to be a difficult task that would require disentangling of environmental from genetic sources of variation in g . Nevertheless, one thing can be concluded from all the evidence gathered so far, is that general cognitive ability is a trait substantially influenced by genes. Even if a more conservative figure of 0.50 is accepted as an estimate of heritability, it still makes g one of the most heritable dimensions of human behaviour and a reasonable target for genetic research.

1.5.3 Relating a trait to variations in DNA sequence

Having demonstrated that a trait has a genetic component, the next challenge is to identify the genetic loci involved and the way in which these determine the phenotype. Broadly speaking, there are two ways of doing this: linkage studies and association studies. Both have advantages and limitations in respect of their application to complex traits. Before considering each individually, some general points on strategies for mapping complex traits will be addressed.

1.5.3.1 Molecular genetic approaches

There are three main approaches to mapping genetic variants involved in a disease/trait: functional cloning, the candidate gene strategy and positional cloning ⁹⁸.

Functional cloning is based on the information about the function of an unknown gene and identification of an underlying protein leads to localisation of the gene. This approach has been useful in only few cases such as cloning of cDNA for Fanconi's anaemia (FACC gene) ⁹⁹ and human coagulation factor IX (F9 gene) ¹⁰⁰.

A candidate gene can be suggested without any prior knowledge of its chromosomal location. Candidates are selected on the basis of phenotypic resemblance with a trait for which the gene is known. Also, if the molecular pathogenesis is known it might suggest that the gene is the member of a known gene family ⁷⁰.

Finally, in positional cloning the general approach is to try to isolate a gene for which only a subchromosomal position is known. Such initial localisation usually comes from linkage and association studies as well as studies looking at chromosomal abnormalities ⁷⁰. Positional cloning is done by constructing detailed genetic and physical

maps which should narrow down the chromosomal region of interest followed by selection of genes likely to be involved in the trait.

1.5.3.2 Increasing the odds for successful mapping

In searching for the genes for complex traits, the main objective is to identify the genetic variation which truly influences the phenotype, or more often, a genetic marker that is correlated with the variant which influences the phenotype. When setting up association and linkage studies one is usually trying to minimise environmental effects by selecting ethnically and culturally homogenous groups. Such populations are not going to be exceptionally homogenous, but the selection is likely to reduce residual genetic effects that are unrelated to the disease and sometimes even genetic heterogeneity which is a common characteristic of complex traits ⁴.

Another approach that can be considered is the selection of samples from the extremes of phenotypic distribution (quantitative traits), or examination of very specific phenotypes. This should result in enrichment of the sample with trait-increasing alleles, thereby increasing the power for detection of the locus ^{101, 102}.

Additionally, increasing the number of marker loci studied increases the probability that one of them will be in significant association with a trait and therefore in strong linkage with a nearby susceptibility gene ⁴. Genetic markers and choices regarding them are discussed later.

1.5.3.3 Linkage studies

In linkage studies related individuals, either siblings or extended pedigrees, are studied in an attempt to localise genes influencing a trait. This is done by examining the co-segregation of the phenotype with genetic markers ⁶².

If two markers are at loci that are close together they may exhibit linkage. This is because it is highly unlikely that a crossover will occur between them at meiosis resulting in a particular haplotype being passed on intact. The proportion of meioses in which recombination occurs between two markers is called the recombination fraction (θ). If the two loci are distant, or on different chromosomes, θ will equal 0.5 ¹⁰³. In other words, their alleles will assort independently, according to Mendel's second law. Linked loci will show departures from independent assortment and will have recombination fraction of less than 50%.

To find out how strong the evidence for linkage is, statistical analysis is performed (Figure No. 15). It is based on a likelihood ratio, the ratio between the likelihood assuming linkage at a given value of θ (L_θ) and the likelihood assuming non-linkage ($L_{0.5}$). The logarithm to the base 10 of that ratio is the LOD (logarithm of odds) ¹⁰⁴.

$$\text{LOD}(\theta) = \log_{10} (L_\theta / L_{0.5})$$

Figure No. 15: Formula for calculating LOD score

The traditional criterion for accepting evidence for linkage is the LOD >3. This value was chosen on the basis of calculations for single gene disorders that if such a threshold is used, 95% of LOD >3 would reflect true linkage ⁷². If multiple genetic

models are tested, the weight of a given LOD score should accordingly be reduced. For this reason, LOD scores between 3 and 6 are considered nowadays as suggestive rather than a definite evidence of linkage for complex traits ¹⁰⁵.

One situation that may be difficult to deal with is a case when a different gene influences a trait in different families - a phenomenon called locus heterogeneity. For this reason, linkage for very heterogeneous traits is usually established in single large families ¹⁰⁴.

Classical methods of linkage analysis require detailed knowledge of the underlying genetic model. For complex traits, specification of parameters essential for analysis can be problematic, leading to reduced power and spurious results ¹⁰⁶. Several non-parametric methods have been proposed as an alternative to classical linkage studies to avoid the problem of mis-specifying parameters. The most commonly used non-parametric methods examine sibling or other relative pairs for increased allele-sharing ¹⁰⁵. Although more appropriate and convenient for studying complex characters, all non-parametric methods are less powerful than the classic parametric ones and require larger sample sizes dependent on the number of genes involved, trait/disease prevalence and heritability of the trait ¹⁰⁶.

Since linkage studies rely on the analysis of related subjects they are only capable of locating a relatively large chromosomal region. With complex traits, linkage peaks may span regions much greater than 10-20 cM, with an inconveniently large number of genes underneath ¹⁰⁷. This means that this type of study may not provide the necessary resolution for locating genes underlying phenotypic variation, but it can be used for initial localisation, especially in cases where there is no prior information on what kind of gene/genes might be involved in influencing a particular trait.

1.5.3.4 Association studies

Association is a phenomenon whereby two features occur together with a frequency greater than expected on the basis of chance alone ¹⁰⁸. Association studies focus on population allele frequencies and test whether a trait and an allele show correlated occurrence in a population.

Genetic loci implicated in complex traits are likely to be of small/modest effect size ⁶³. The linkage method, although successful in detecting major genes, has limited power to detect genes of modest effect. For loci conferring genotypic relative risks (GRR) of 2 or less, the sample sizes required are generally beyond reach (> ~2500 families) ¹⁰⁹. On the other hand, the sample sizes required for association studies are much smaller, even for detecting smaller signals with GRR of 1.5, and are generally less than 1000 affected individuals ¹⁰⁹. So, association is a powerful technique and is increasingly seen as the method of choice for mapping complex trait loci.

The measure of association can be qualitative (presence or absence of a significant association) or quantitative (strength of association). The latter one can be evaluated by several statistical parameters such as: risk difference, relative risk, odds ratio and attributable risk ¹⁰⁸. The χ^2 test detects the existence of a significant association.

The basic approach of the method is to identify marker loci at which some alleles are more frequent (or deficient) among individuals with a particular phenotype than among controls. A statistically significant difference (positive association) can arise for several reasons ⁶²:

- A) if the marker itself is contributing to overall phenotypic variance (pure or direct association),
- B) if there is a tight physical linkage between a marker and a trait locus (linkage disequilibrium or indirect association), or

C) if the association observed results from population admixture and an inadequately matched control group (spurious association).

Spurious associations are a big problem in population-based studies. To significantly reduce the number of such, it is crucial to select suitable controls. It is of utmost importance that the two groups compared are matched for ethnicity to avoid any false associations that have nothing to do with the examined trait but rather reflect ethnic differences in marker allele frequencies.

Family-based association methods help overcome this problem by using 'internal' controls ¹⁰⁸. The two most frequently used are the *Haplotype Relative Risk (HRR)* ¹¹⁰ and the *Transmission Disequilibrium Test (TDT)* ¹¹¹. In the HRR allele frequencies in affected offspring are compared with frequencies in a "control group" consisting of non-transmitted parental alleles. In the TDT, the frequency of transmission of alleles from heterozygous parents to affected offspring is examined.

In addition to being more costly than case-control studies, family-based studies have to face problems in relation to sample collection. Diseases with a later age of onset are especially unfavourable to the family based design, because the parents are often dead and/or other members of the nuclear family are unreachable.

A further argument to be considered in favour of case-control studies is that the effect of population stratification is actually negligible within an ethnic group, even if defined vaguely as Caucasian, Black, Asian or Hispanic ¹¹². So, unless an investigator is very careless and mixes different ethnic groups with different disease prevalence and marker frequencies, spurious associations due to population stratification should be extremely rare ¹¹².

One easy and cost effective way to control for stratification in population based studies was proposed by Pritchard ⁶. As allele frequencies at random marker loci may differ among ethnic groups ¹¹³, one should expect to find allele frequency differences between such groups throughout the genome. If cases and controls are well matched,

then significant allele frequency differences should be observed only in close proximity to a disease susceptibility allele. In other words, if stratification due to different ethnic representation in case and control groups is present, then markers unlinked to a disease locus should also show associations with the phenotype. It has therefore been suggested that case-control based studies should include in their design 15-20 randomly chosen microsatellite markers, or 30 biallelic markers in order to keep the overall Type I error rate in reporting associations to $\leq 5\%$ ⁶.

Another common problem which may lead to false positive results is the application of insufficiently rigorous statistical analysis. This is often due to inadequate correction for the number of questions asked/markers tested. Since association studies of complex traits are likely to involve testing a large number of markers, correcting the final p value using the Bonferroni correction method ¹¹⁴ will require a statistical significance several orders of magnitude greater in order to remain significant. Such levels of significance are almost impossible to reach, and applying such a conservative multiple-test correction could result in real signals being missed. Approaches such as consistent replication of positive findings on an independent sample have been suggested as an alternative yet powerful way of circumventing the problem ¹¹⁵.

An important limitation to be considered in association studies of complex traits is heterogeneity. Many different environmental and genetic factors involved may result in a difficulty to find a single genetic marker associated with the trait in question within the practical scale of a study ¹¹⁶.

Even with all these limitations, association studies are still one of the easiest ways of filtering out genes which may influence a trait of interest, assuming all precautionary measures are taken. Also, they are frequently the inevitable choice if the character studied is complex in nature and is likely to be influenced by a number of genes in combination with environment.

1.5.4 Linkage disequilibrium (LD) mapping

Association studies can either be direct or indirect ¹¹⁷. Direct studies rely on studying variants in coding and regulatory regions of genes in the hope that these will include variants contributing directly to a particular trait. Indirect studies rely on association between a trait and a marker allele located close to and in *linkage disequilibrium (LD)* with an allele contributing to phenotypic variance. In other words, even variants that are not included in the screen would be assayed indirectly through LD with nearby markers. This approach is known as *LD mapping* ¹¹⁸.

LD mapping relies on the assumption that a single ancestral mutation is responsible for a large proportion of disease carriers (individuals with a certain trait) in a present-day population ¹¹⁹. The mutation arises on a particular ancestral haplotype which slowly degrades by recombination with passing generations. However, this haplotype should largely be preserved within the region around the mutation because loci in close proximity to each other will rarely be separated by recombination events. The approach makes use of the large number of generations since the first appearance of the mutation and allows substantial narrowing of the interval in which a disease gene might lie.

1.5.4.1 Measures of LD

Two alleles are said to be in linkage disequilibrium, also called allelic association, if they occur together at a frequency greater than expected if the alleles were segregating at random. In other words, the frequency at which these alleles occur together will depart from the simple product of their allele frequencies, $p_A \times p_B$, where p_A is the frequency of allele A at the first locus and p_B the frequency of allele B at the second

locus ¹²⁰. This means that unrelated people who have a certain allele at one locus will have statistically greater chance of having a particular allele at a second locus if these two loci are in LD.

One of the earliest measures of disequilibrium, symbolised by D , was proposed nearly forty years ago ¹²¹. It quantified LD as a difference between expected and observed haplotype frequencies for two loci in question. Because its numerical value is dependent on allele frequencies, D cannot be used as a measure of LD strength or for comparing levels of LD ¹¹⁸.

In addition to D , there are two other commonly used measures of LD: D' and r^2 . D' is determined by dividing D by its maximum possible value. $D' = 1$ when LD is complete. Intermediate values, $D' < 1$, tend to be biased upward inversely with sample size ¹²² and should not be used for comparisons of the strength of LD between studies

¹¹⁸.

$$\begin{aligned}
 D &= p_{AB} - p_A p_B \\
 D' &= D / D_{max} \\
 r^2 &= D^2 / p_A p_a p_B p_b
 \end{aligned}$$

Figure No. 16: Calculating different measures of LD.

r^2 , the correlation of alleles at the two sites, is formed by dividing D^2 by the product of the four allele frequencies at the two sites. Although it is allele frequency and sample size dependent, r^2 is much less inflated in small samples than D' ¹²², and has been proposed as the measure of choice for quantifying and comparing LD levels ¹²³. While

$r^2 = 1$ is known as perfect LD, values of r^2 above 1/3 might indicate sufficiently strong LD to be useful for mapping ¹¹⁸.

In addition to pairwise measures of LD (*i.e.* D , D' , r^2), the population recombination parameter, $4N_e r$, facilitates comparisons between regions ¹¹⁸. However, estimation of this parameter from genotyping data is computationally very challenging and is currently a subject of intense scientific research ¹²⁴.

1.5.4.2 What is the extent of LD?

Linkage disequilibrium is not a strict monotonic function of distance. Both strong LD between very distant markers and virtually non-existent LD between physically close markers are readily observed in the human genome. Mutation and recombination might have the most evident impact on linkage disequilibrium, but other factors are likely to further shape the extent of LD and change the simple relationship between the strength of LD and the physical distance. Some of these are:

- A) *Genetic drift*: This causes changes in haplotype frequencies due to random sampling of gametes. The effect is especially strong in small and stable populations, where low frequency haplotypes tend to be lost, resulting in increased LD ¹²⁵.

- B) *Admixture*: This is the phenomenon whereby two or more genetically distinct populations admix and create a so-called "hybrid" population. In such populations that have undergone recent admixture very strong LD is observed over large distances. The speed of LD decay will depend on the size of the hybrid population,

and whether it experiences a constant external influx from one of its "originator" populations (slow LD decay), or has become isolated (steep LD decay) ¹²⁶.

- C) *Population growth*: Rapid population growth causes faster LD decay through reducing the effect of genetic drift.
- D) *Population bottlenecks*: Bottlenecks result in substantial reduction in population size and haplotype diversity leading to increase in the extent of LD ¹²⁷.
- E) *Sub-populations and inbreeding*: Tend to increase LD.
- F) *Natural selection*: The entire haplotype flanking the favoured variant becomes highly frequent in the population, a phenomenon called "hitchhiking effect". Negative selection, although milder in effect, can also strengthen LD by removing haplotypes with deleterious variants.
- G) *Variable recombination and mutation rates*: There will be little or no LD between loci within regions with high recombination rate, so-called "hot spots", and within those characterised by higher mutation rate, like CpG islands ¹²⁸.
- H) *Gene conversion*: This is the phenomenon whereby a short stretch of one copy of chromosome is transferred to the other copy during meiosis. The process is equivalent to two very closely spaced recombination events ¹¹⁸, and results in rapid LD breakdown ^{129, 130}.

So, depending on the size and history of the population, LD is seen over varying distances from a trait locus. Haplotype signatures tend to be shorter in older outbred populations ¹²⁷ than in younger population isolates ¹³¹. Recent computer based

simulations predicted that sufficient LD for indirect association studies is unlikely to extend beyond an average distance of 3 kb in the general population,¹¹⁷. The 3 kb figure was met with a widespread scepticism^{132, 133, 127, 128}.

Collins (1999) predicted that in large outbred populations disequilibrium should be detectable within 100-300 kb (0.1 – 0.3cM) of a trait locus after analysing data from 1000 pairs of loci¹³². In a similar attempt to throw some light on extent and distribution of LD, Abecasis (2001) examined 127 polymorphisms in a sample of 575 chromosomes from unrelated individuals of British ancestry and found out that for half of the markers LD extends to 50 kb¹³³.

LD may extend over much larger distances (>5 cM). This is particularly true for small population isolates^{134, 131}, making such populations attractive for whole genome scans for association with complex traits. When talking population isolates, one usually thinks of populations coming from Finland, Iceland, Sardinia and Japan. However, these populations show levels of LD (<1 cM) similar to other populations in Europe. This is likely to be due to combination of two factors: the relatively large number of founders and the growth these populations experienced since their foundation¹³⁵. By contrast, small populations that have remained stable in size over long periods, like Scandinavian Saami¹³⁶, are more likely to have experienced much stronger effect of genetic drift which resulted in LD extending over larger regions (>10 cM)^{125, 137}. Shifman and Darvasi (2001) argue that apparent similarity in LD between outbred populations and population isolates (*i.e.* Finland and Iceland) is true only for SNPs separated by up to 200 kb, while those separated by more than 200 kb show significant differences in LD level between these populations¹³⁸.

Excess of LD is also observed in admixed populations. Lemba, a hybrid population generated by Bantu-Semitic admixture, is a fine example of one. Significant LD is present between markers ≤ 21 cM apart¹³⁹. Pfaff *et al.* reported similar findings in two African American populations where LD extended over 10 cM¹²⁶.

The attractiveness of populations that show extensive linkage disequilibrium lies mainly in the fact that genome-wide mapping would be achievable with a smaller set of already available markers from public databases, but one has to bear in mind that the resolution of such mapping will be relatively low ¹²⁷.

1.5.4.3 Block structure of LD

Several recent studies ^{140, 141, 142, 143, 144} indicate that linkage disequilibrium in the human genome is highly structured into blocks of low haplotype diversity separated by recombination hot-spots (Figure No. 17) ¹.

Haplotype blocks, characterised by extensive LD, can span from only a few to >100 kb. Even though most (~94%) of all crossovers were found to lie within recombination hot spots, resulting in a rapid LD breakdown, they are not exclusive to these short stretches of chromosomes (1 to 2 kb) ¹⁴². This may explain a within-block decay of LD which, if present, is usually much more gradual than that in hot spot regions.

If there is very little recombination between the blocks, a high inter-block correlation called long-range LD, will be present. The phenomenon is characterised by lower haplotype diversity where a small number of haplotypes observed over longer stretches of chromosomes account for a large proportion of chromosomes in a given sample ¹⁴¹.

Since haplotype structure is highly dependent on population history, different patterns of LD are likely to be observed in different populations and different chromosomal regions. In some extreme cases of large and stable populations, a sufficient number of crossover events may have occurred within blocks for the clear block-like structure of LD to be completely absent ¹. The lower LD seen in the African population ¹²⁷ could therefore be a result of sharper decay of disequilibrium within and/or between blocks.

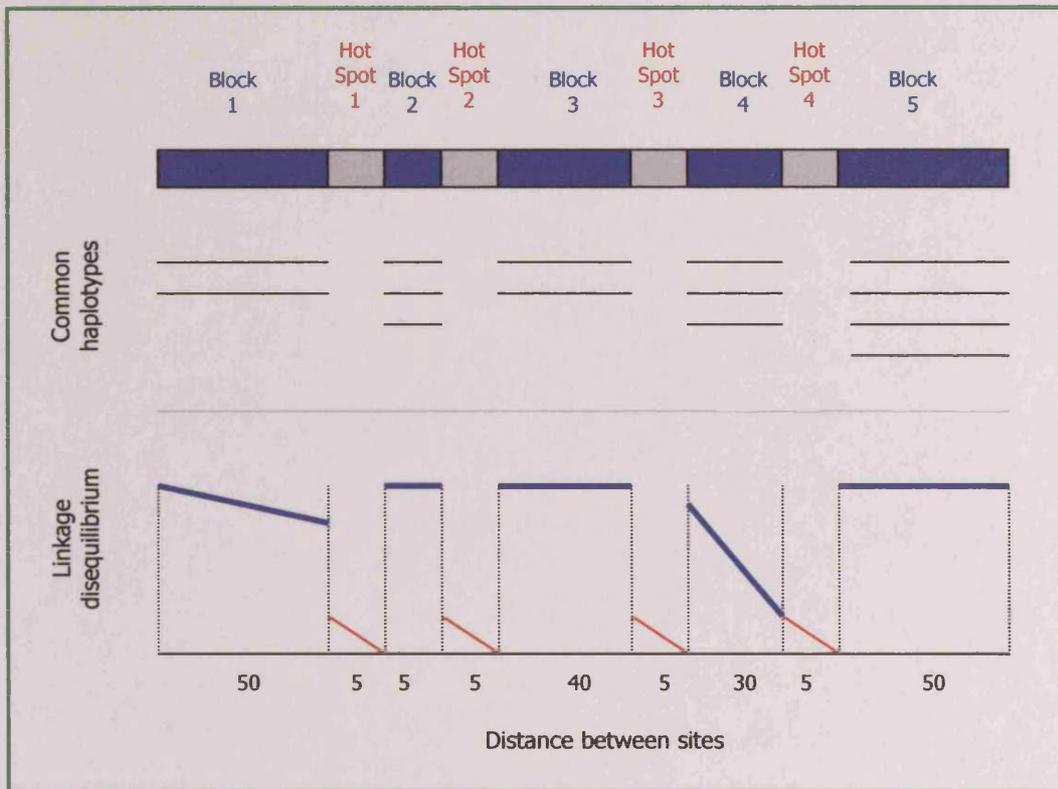


Figure No. 17: Islands of LD ¹

If the block-like structure of the human genome proves to be a rule rather than exception, it is likely to have a profound impact on association studies. The number of SNPs one will have to genotype could drastically be reduced to only 2-5 per domain, bringing down the total number of SNPs necessary for a genome-wide scan to 80,000-200,000 ¹⁴². For variants located within hot spot clusters this number will have to be higher.

More recently, a term linkage disequilibrium unit (LDU) was introduced ¹⁴⁵. One LDU corresponds to one swept radius which is defined as the average extent of useful LD ¹⁴⁶. Regions of extensive LD have fewer LD units while regions with high levels of recombination have greater number of LDUs. Based on chromosome 22 data, Tapper

et al. estimate a total of ~50,000 LDUs per human genome and anticipate that a number of SNPs required for an efficient genome screen will be a multiple of this number (200,000 – 300,000).

Although attractive in terms of resources necessary for carrying out an association study, the block structure of LD might mean that the maximum resolution that can be achieved with LD mapping cannot go much further than ~100kb^{127, 146}. If finer resolution is desired, populations with less extensive LD may have to be considered.

1.5.4.4 Power to detect LD

The power to detect LD is affected by the method used, the number of available samples, the recombination and mutation pattern of the region, the degree of locus and allele heterogeneity, age of the mutation, the number and type of markers used, and population history¹³⁷.

Long and Langley (1999) estimate that sample sizes of ~500 individuals typed with ~20 SNPs spaced throughout a candidate gene region (~250 kb, if one assumes that one 4Nc in humans is equal to 5 kb and that there are 50 such units in a candidate region) should provide sufficient power to detect QTLs that account for ~5% of the variation in a complex trait¹²⁴. Results from simulation studies they carried out also suggest that a greater increase in power will be achieved by increasing sample sizes than by increasing the number of typed SNPs. That is particularly true for smaller regions. Increasing the number of SNPs will be helpful only if recombinationally larger regions are considered.

Allelic and locus heterogeneity can result in marked decrease in power^{147, 4}. Increased allelic heterogeneity gives rise to a greater number of possible haplotype backgrounds each harbouring a particular disease locus variant. This dilutes the signal

coming from any one of these haplotypes, resulting in less LD and lower power to detect it ⁴. The chances of detecting LD can also be reduced if the allele conferring susceptibility is common relative to the population size ^{125, 108}.

The length of time since the disease causing mutation occurred is also important. Mutations responsible for complex diseases tend to be older because they are typically not subject to strong natural selection. Older mutations allow greater resolution with LD mapping studies, but being too old might result in equilibrium already being reached and/or many new mutations accumulated ¹⁴⁷.

Because of its short range in large outbred populations, LD mapping should preferably be used as a method for fine mapping once a region of interest has been found by linkage analysis - the strategy successfully applied for mapping complex disease loci such as a Crohn's disease susceptibility gene ¹⁴⁸, susceptibility genes for asthma ^{149, 150}, ulcerative colitis ¹⁵¹, diabetes mellitus type 2 ¹⁵², autism ¹⁵³ and schizophrenia ¹⁵⁴.

1.5.5 Choice of markers

Genetic markers are heritable characters (loci) with multiple states (alleles) at each character ¹⁵⁵.

Microsatellites, or Short Tandem Repeats (STRs) became the markers of choice for genetic mapping studies in 1989 ¹⁵⁶. Their main feature is the presence of many alleles and high levels of heterozygosity which maximises the number of informative meiotic events essential for resolving the segregation of alleles through a family. Such markers provide an ideal tool for pedigree-based linkage analysis ¹⁰⁶. Three types commonly

used are recognised on the basis of the number of nucleotides comprising each repeat unit. Those are di-nucleotides, the most abundant, tri- and tetra-nucleotides.

Recent developments in mutation detection technology led to emergence of a new generation of markers called Single Nucleotide Polymorphisms (SNP). These markers are usually biallelic and less polymorphic than microsatellites, but they have the advantage of being highly abundant in the genome and, because of their binary nature, favourable for high-throughput automated genotyping technologies.

The power to detect linkage between a marker and a disease locus and the accuracy of the estimation of the map location of the disease gene greatly, but not exclusively, depends on the right choice of markers. In general, if the frequency of the associated allele is unknown and the distance between the marker and the disease locus is relatively large, a microsatellite marker would be a better choice for population-based studies¹⁵⁶. Sham *et al.* have shown that highly polymorphic markers also have greater power in cases where there are multiple ancestral disease mutations and suggested ten alleles to be a sufficient allele number for detection of LD¹⁵⁷. Nevertheless, it has been suggested that when choosing between the markers with the same heterozygosity one should opt for the one with fewer alleles¹⁵⁸.

However, SNPs are likely to outperform microsatellites in LD mapping studies where the distance between the marker and the disease locus is small, so that the microsatellite mutation rate is not negligible compared with the recombination rate¹⁵⁶. Typing a large number of SNPs overcomes the problem of the low informativeness that an individual SNP suffers from and offers the opportunity to further increase power by generating haplotype information in a small genomic region of interest^{156, 159}. On the other hand, if SNPs are in strong linkage disequilibrium with one another less information and less power will be contributed¹³⁷.

The use of biallelic markers rather than multiallelic microsatellites for LD testing in genome screens requires much denser distribution of markers to achieve the same power¹⁵⁸. Apart from the feasibility question one has to consider the increase in Type I

error related to a larger number of markers tested. Even more problematic is the fact that linkage disequilibrium does not necessarily have to exist even for SNPs within the disease gene ¹⁰⁶. An increasing number of studies are trying to deal with these issues through investigating the application of haplotype analysis to LD mapping. So far, promising results have been reported for single-gene disease models ¹⁵⁹, whereas complex traits and diseases are still to be considered.

Linkage analysis with microsatellites used to identify regions of interest followed by association-based analysis employing SNPs to narrow down the region to a size manageable for positional cloning is a strategy recommended by Almasy and Blangero ¹⁰⁶. Many research groups have adopted such approach and are focusing on SNP genotyping of linkage peaks previously detected with microsatellites ^{149-151, 154}.

However, more recent reports suggest that microsatellite linkage mapping may soon become a strategy of the past. Direct comparisons of linkage peaks obtained through the use of microsatellite markers and densely distributed SNP indicated a high degree of correspondence between the two approaches, but there were instances where more traditional microsatellite approaches had insufficient power to detect highly significant linkage peaks ^{160, 161}. In addition, high-density SNP scans offer considerable savings in downstream mapping owing to better definition of the linkage peaks ¹⁶⁰.

1.6 Mapping studies on intelligence

Although a number of studies offer support for a relatively large genetic contribution to variance in *g*, there have been a few attempts to identify the specific genes.

The first mapping study on intelligence focused on available DNA markers, mostly RFLPs (restriction fragment length polymorphisms), in or near genes likely to be of neurological relevance^{162, 163}. 100 markers were examined for allelic association by looking at allele frequency differences between high-IQ (IQ > 130) and low-IQ groups (IQ < 82). Despite a two-stage study design (original and replication stage) and selection of individuals from extremes of the phenotypic distribution, small sample sizes offered limited statistical power to detect genes of small effect size. Several significant associations ($p < 0.05$) were found in the first stage of the study. Two of them, alcohol dehydrogenase 5 (ADH5) and nerve growth factor beta polypeptide (NGFB) showed allele frequency differences in the same direction in the replication sample but failed to reach statistical significance.

A more systematic approach was taken by Chorney *et al.* (1998) who, instead of looking at candidate genes, used a dense map of microsatellite markers on chromosome 6p to look for allelic association with general cognitive ability¹⁶⁴. The authors reported an association with a marker situated within the gene for insulin-like growth factor 2 receptor (IGF2R), a finding later replicated on a larger sample and different polymorphism in IGF2R¹⁶⁵.

With fast development of new molecular techniques and increased public availability of markers, both microsatellites and SNPs, it would be reasonable to expect further mapping efforts aimed at detecting some of the QTLs responsible for phenotypic variance in *g*.

1.7 Animal research

Cognitive ability, especially learning, memory and problem solving, is also studied in other species such as rat, mouse, marine snail *Aplysia* and fruit fly. In one of the first

such experiments during 1920s, rats were selectively bred for their performance in learning the maze in order to find food ¹⁶⁶. During the 1950s and 1960s studies of inbred strains of mice showed that genes are important contributors to learning and memory. These maze-dull and maze-bright rats were also used in a study of genotype-environment interaction ¹⁶⁷, the only research on learning, so far, that managed to show differential response of genotypes to environment ¹⁶⁸. Late 1970s were characterised by introduction of *Aplysia* for the study of learning and memory. This simple organism allowed relatively easy mapping of learned behaviour to synaptic change ¹⁶⁹. Even though mammalian brain is much more complex, *Aplysia* provided an important example of the relationship between learning and plasticity, as well as elucidation of various mechanisms of plasticity that are conserved across the species ¹⁷⁰.

Candidate gene studies using knock-out strains of mammals have generated a substantial amount of data on the involvement of various genes and their products in particular aspects of mental functioning. An important contribution to knock-out studies was made by Tsien *et al.* who developed a method to create mice in which the deletion of almost any gene will be restricted to specific brain region such as the hippocampal CA1 region ¹⁷¹. Creation of such knockouts allows more precise analysis of the impact of specific gene mutations on animal behaviour.

A substantial proportion of current animal research focuses on investigation of molecules involved in neuronal migration, differentiation, axon outgrowth and synapse formation, with particular focus on their role in activity-dependent synaptic plasticity. Integrin-deficient ¹⁷², neural cell adhesion molecule (NCAM)-deficient ¹⁷³ and LIM kinase 1-deficient ¹⁷⁴ mice are just some of the many strains that provided strong evidence for involvement of such molecules not only in synapse architecture and development, but also in mediation and regulation of synaptic plasticity.

Another vigorously explored avenue is directed towards examining receptors that may have substantial impact on processes leading to changes in synaptic transmission and, by doing so, on learning and memory. The importance of AMPA and NMDA

receptors has been acknowledged for a long time, but it was the production of mutant mice expressing different receptor subunits that provided more insight into their roles in learning and memory. GluR2 subunit of AMPA receptor is correlated with lower calcium ion permeability ¹⁷⁵, and not surprisingly GluR2 mutant mice were found to exhibit enhanced levels of synaptic potentiation ¹⁷⁶. A few years later, mice with enhanced learning ability and over-expressing the 2B subunit of NMDA receptors were generated ¹⁷⁷.

In addition to these two most abundant receptor groups, other receptors were also found to have an impact on behavioural processes, particularly spatial learning. Opioid ¹⁷⁸, interleukin-1 ¹⁷⁹ and oestrogen receptor knockouts ¹⁸⁰ all show disturbed synaptic plasticity and profound deficits in spatial ability.

Such studies are throwing more light on complex processes of learning-related synaptic plasticity and are bringing our attention to genes likely to be implicated. Results reported by many research groups have confirmed some pre-existing hypotheses regarding molecular mechanisms that take place during synaptic activity. Equally, if not more, important is the contribution made by groups that brought our attention to novel mechanisms that appear not to be alternative, but rather complementary to already known ones.

2 AIMS OF THE STUDY

General cognitive ability is one of the most heritable human behavioural traits^{88, 89}, and a good candidate for molecular genetic research.

This study will attempt to identify some of the genetic loci responsible for overall variance in g . Two approaches will be used: 1) a QTL association study with a dense map of microsatellite markers (aiming for a 1cM grid), and 2) an investigation of functional candidate genes.

3 GENERAL MATERIALS AND METHODS

3.1 *Sample*

The project includes three samples: 1) an original sample of 101 high *g* cases and 101 average *g* controls, 2) a replication sample of 100 very high *g* cases and 100 average *g* controls, and 3) a within-family replication sample of 196 parent-offspring trios in which children have high *g* scores. The third sample was available only to our project collaborators in London.

The samples were obtained from areas around Cleveland (Ohio, USA). Subjects were restricted to non-Hispanic, white children in order to reduce Error 1 rate due to population stratification. Subjects were excluded if one of the maternal or paternal grand-parents was not white. Informed consent was obtained from all participants. For all subjects in the original case-control groups, permanent cell lines were established from blood samples. DNA for the remaining subjects was extracted from cheek swabs.

3.1.1 Original sample

The original controls were selected for IQ scores between 90 and 110 (mean IQ 102.2; 6.3 SD) from a larger group of children assessed using the Wechsler Intelligence Scale for Children Revised (WISC-R) ¹⁸¹. Average age was 13.

First 51 children of the original cases group were selected for IQs above 130 from the same group of children (mean IQ 136.0; 9.3 SD). Average age was 10. The remaining 50 original high *g* individuals were selected from the highest-scoring

individuals from the Study of Mathematically Precocious Youth (SMPY) ¹⁸² that sat the Scholastic Aptitude Test (SAT, college entrance exam) four years early. These children obtained SAT scores 4 standard deviations above the mean (equivalent to IQ scores > 160).

3.1.2 Replication sample

Replication samples were obtained from the same Cleveland area and in a similar fashion as the original group. The controls had an average IQ of 102.4 (7.6 SD), but unlike the original group, the subjects were adults (33 years average age). They were assessed by the Wechsler Adult Intelligence Scale-III ¹⁸³.

The replication high *g* group was selected from the remaining highest-scoring individuals in the SMPY sample. IQs estimated from their SAT results were > 160.

3.1.3 Parent-offspring trios

The sample of 196 parent-offspring trios was obtained from the SMPY study. High scoring individuals were selected on a basis of availability of their parents. About one third of the children forming this group were also used as replication cases for the study.

3.2 Creation of DNA pools

DNA pools were constructed from individual DNA samples that were brought down to 6ng/μl prior to pooling, making sure that every individual contribute the same amount of DNA. The volumes of the pools were determined from the number of markers planned to be tested and from the amount of DNA that is used per standard PCR reaction. All pools were constructed in triplicate.

DNA quantification prior to pooling was performed at three concentration levels (~80ng/μl, ~20ng/μl and at 6ng/μl) using the PicoGreen™ fluorescent assay. PicoGreen dye was added to genomic DNA samples that were diluted in 1xTE buffer (1:20 or 1:50 dilution factor depending on the expected DNA concentration) and pre-aliquoted into White Cliniplates™ (Labsystems). After incubation of 4min at room temperature samples were analysed by a Flourosan Ascent fluorometer. The method gives more reliable readings than the simple spectrophotometry because it relies on the dsDNA intercalating property of the dye. This ensures that only double stranded DNA molecules generate signal that is translated by the Ascent software into a DNA concentration of the tested sample. However, presence of some compounds that commonly contaminate nucleic acid preparations, like phenol, ethanol and chloroform, can increase signal intensity by 13%, 12% and 14% respectively (figures adopted from Molecular Probes product information sheet).

3.3 Polymerase Chain Reaction (PCR)

The polymerase chain reaction (PCR) is a technique for rapid amplification of a target DNA sequence *in vitro*.

The DNA fragment to be amplified is determined by selecting primers. Primers are short oligonucleotide strands (usually 18-25 bp) that are complementary to the beginning (forward primer complementary to the "minus" strand) and end (reverse primer complementary to the "plus" strand) of the DNA fragment to be amplified. The PCR reaction is carried out in a thermal cycler which heats and cools the reaction mixture to the precise temperature and length of time required for each step of the reaction.

PCR primers for microsatellite markers used in the Genome Scan study were designed by their original submitters (<http://cedar.genetics.soton.ac.uk/>), and were purchased from MWG-Biotech (Germany). Primers for candidate gene studies were designed using the Whitehead Institute for Biomedical Research software *Primer 3* (http://frodo.wi.mit.edu/cgi-bin/primer3/primer3_www.cgi) and purchased from Sigma-Genosys.

3.3.1 Standard PCR

Standard PCR reaction mixture and thermal-cycling protocol were used for candidate gene studies. Annealing temperatures and optimal Mg²⁺ concentrations were deduced empirically from optimisation reactions. Main features of a standard PCR are presented in Tables 3 and 4.

Table No. 3: PCR reaction mixture preparation. All values apply for ONE reaction.

PCR mixture component	Volume (μ l)
DNA (6 ng/ μ l)	4.00
H ₂ O	4.96
MgCl ₂ (25 mM)	0.24
Reaction buffer (x10)	1.20
dNTPs mix (5 mM each)	0.96
Forward primer (5 pmol/ μ l)	0.28
Reverse primer (5 pmol/ μ l)	0.28
<i>Taq</i> polymerase	0.06

Table No. 4: Thermal cycling protocol for a STANDARD PCR reaction.

Cycling Step	Temperature	Time	No of cycles
Denaturation	94°	5 min	1
Denaturation	94°	30 sec	35
Annealing	T° deduced from optimisation experiments	30 sec	
Extension	72°	45 sec	
Final Extension	72°	10 min	1

3.3.2 Touchdown PCR

In addition to standard, a Touchdown protocol was used. Here, a template DNA is amplified using a range of annealing temperatures in a single reaction. Touchdown PCR simplifies the process of determining optimal primer annealing temperatures and is particularly useful for studies investigating a large number of genetic markers. It is also helpful in cases where a non-specific primer binding is suspected¹⁸⁴.

During the initial cycles of touchdown PCR, annealing takes place at temperatures 5° to 10°C above the calculated T_m. In subsequent cycles, the annealing temperature is gradually reduced by 1-2°C each cycle until it has reached approximately 5°C below the calculated T_m. The touchdown thermal cycling protocol successfully employed for

genotyping of more than a thousand microsatellite markers in the Genome wide screen for association with IQ (Chapter 4) is presented in Table 5.

Cycling Step	Temperature	Time	No of cycles
Denaturation	94°	5 min	1
Denaturation	94°	30 sec	10
Annealing	60° -1° with each additional cycle	30 sec	
Extension	72°	45 sec	
Denaturation	94°	30 sec	25
Annealing	50°	30 sec	
Extension	72°	45 sec	
Final Extension	72°	10 min	1

Table No. 5: Thermal cycling protocol for a TOUCHDOWN (60°-50°C) PCR.

3.3.3 Post-PCR product treatment

3.3.3.1 Klenow fragment treatment

Post-PCR treatment of amplified products with *E. coli* DNA polymerase I Klenow fragment (Amersham) to repair 3' overhanging ends and generate blunt-ended products was used for the pooled genotyping stage of the genome-wide screen with microsatellites for the purpose of reducing the stutter band effect (for detailed discussion see Chapter 4).

Mixture component	Volume (μl)
Pooled PCR products	12.5
H ₂ O	0.6
Reaction buffer (x10)	1.5
Klenow fragment (5u/ μ l)	0.4

Table No. 6:

Components of a "Klenow mixture" for post-PCR treatment of amplified products. Reaction mixture is incubated at 30° for 90 minutes.

3.3.3.2 *Exonuclease I* and Shrimp alkaline phosphatase treatment

Certain techniques, such as DNA sequencing and SNaPshot genotyping (described further down), require PCR products to be free from unincorporated primers and dNTPs. Purification was achieved by adding 2U of *Exonuclease I (ExoI)* (Amersham) and 2U of Shrimp alkaline phosphatase (SAP) (Amersham) to a 12 μ l PCR product. The mixture was incubated at 37°C for 75 min, after which the reaction was stopped by heating the products at 80°C for 15 min.

3.4 ***Agarose Gel Electrophoresis***

Agarose gels provide the most commonly-used means of isolating, purifying and size separating fragments of DNA. The equipment and supplies necessary for conducting agarose gel electrophoresis are relatively simple and include:

- 1) Agarose MS (Boehringer Mannheim) for separation of 50bp – 1500bp fragments
- 2) An electrophoresis chamber and power supply

3) Gel casting trays and well-forming combs

4) Electrophoresis buffer, *i.e.* Tris-borate-EDTA (0.5X TBE, National Diagnostics)

5) Loading buffer which contains a dense substance to allow the sample to "fall" into the sample wells and one or two tracking dyes, which migrate in the gel and allow monitoring of how far the electrophoresis has proceeded. It can be prepared by adding 3.0g Ficoll 400 (BDH), 0.22g EDTA (Sigma), 0.02g Bromophenol Blue (Sigma) and 0.02g Xylenecyanol (Sigma) to 20ml H₂O. The dye mixture is kept over-night in a warm water bath.

6) Ethidium bromide, intercalating fluorescent dye used for staining nucleic acids

7) 1kb DNA ladder

8) Transilluminator (an ultraviolet lightbox) fitted with Polaroid camera

To pour a gel, agarose powder is mixed with electrophoresis buffer to the desired concentration (1.5 – 3.0%), then heated in a microwave oven until completely melted. By using gels with different concentrations of agarose, one can resolve different sizes of DNA fragments. Higher concentrations of agarose facilitate separation of small DNAs, while low agarose concentrations allow resolution of larger DNAs. Ethidium bromide (1µl of 10mg/ml is sufficient for a 100ml gel) is added to the gel at this point to facilitate visualization of DNA after electrophoresis. After cooling the solution to about 60°C, it is poured into a casting tray containing a sample comb and allowed to solidify at room temperature or in a refrigerator. After the gel has solidified, the comb is removed.

DNA samples are mixed with 4µl loading buffer per 12µl sample. 10µl of this mixture are then loaded into the sample wells and a current is applied (~100 volts for 30-90 min depending on the fragment size). DNA migrates towards the anode. The distance DNA has migrated in the gel can be judged by visually monitoring migration of the tracking dyes. Bromophenol blue and xylene cyanol dyes migrate through agarose gels at roughly the same rate as double-stranded DNA fragments of 300 and 4000 bp,

respectively. To visualize DNA, the gel is placed on a UV transilluminator and a photograph is taken.

3.5 Mutation detection

Analysis of candidate genes often involves examination of sequence polymorphisms and their possible associations with a trait investigated. Denaturing High Performance Liquid Chromatography (DHPLC) is a mutation detection method that was used in this study and was accompanied by subsequent sequencing for ultimate confirmation of polymorphisms detected by DHPLC.

3.5.1 Denaturing High Performance Liquid Chromatography (DHPLC)

Denaturing high performance liquid chromatography (DHPLC) analysis is a reverse-phase ion-pair chromatography that allows the identification of single-nucleotide mismatches and small insertion-deletion polymorphisms within an amplified DNA fragment several hundred base pairs in length¹⁸⁵. It is a rapid and semi-automated method based on differential separation and detection of mismatched DNA strands (heteroduplexes). Formation of heteroduplexes is a crucial step. It is performed on crude PCR products which are rapidly heated at 96°C for 2 min to allow melting followed by gradual decreasing sample temperature from 96°C to 56°C over a period of 40 min (temperature decrement of 1°C per minute) to allow re-annealing and enable the formation of heteroduplexes. At the appropriate column temperature, partially melted DNA heteroduplexes, formed after re-annealing of normal and mutant DNA

strands, have relatively reduced affinity to the hydrophobic matrix and are therefore eluted earlier from the column, enabling DNA variants to be detected with high sensitivity.

The method was performed using the Transgenomic WAVE system. Aliquots of 5-7 μ l PCR samples (that underwent melting procedure described above) were loaded onto a preheated reverse-phase column based on non-porous polystyrene-divinyl-benzene particles (DNASep column, Transgenomic). DNA (homoduplexes with or without heteroduplexes) was eluted from the column by a linear acetonitrile gradient in 0.1 mM triethylamine acetate buffer (TEAA, Transgenomic), at a constant flow rate of 0.9 ml/min. DNA was detected at 260 nm. The gradient was formed by mixing buffer A (0.1 mM TEAA) and buffer B (0.1 mM TEAA, 25% v/v acetonitrile). Optimal combination of buffers and the column temperature, to allow fragment retention between 3 and 5 min, depends on a nucleotide composition of each of the fragments analysed. A simple and quick publicly available software based on simulated annealing algorithms was used to obtain optimal parameters for each of the fragments analysed (<http://insertion.stanford.edu/meltdoc.html>).

Each fragment was analysed on 15 high *g* individuals – mutation screening set. The size of the screening set gives sufficient power (95%) to detect polymorphisms with a minor allele frequency of 0.1. For each of the fragments analysed chromatograms were obtained for all the samples in the screening set. Elution curves were compared and a shift in a trace pattern was treated as indicative of the presence of a heteroduplex.

3.5.2 DNA Sequencing

Fragments showing elution curve shifts on DHPLC were subjected to sequencing for confirmation and characterisation of polymorphisms.

DNA sequencing is the determination of the precise sequence of nucleotides in a sample of DNA. The most popular method for doing this is called the dideoxy method. The dideoxy method gets its name from the critical role played by synthetic nucleotides that lack the -OH at the 3' carbon atom. Chain elongation stops once a dideoxynucleotide (ddNTP) is incorporated because there is no 3' -OH for the next nucleotide to be attached to. For this reason, the dideoxy method is also called the chain termination method¹⁸⁶.

Because all four deoxynucleotides are present in a standard sequencing reaction, chain elongation proceeds normally until, by chance, DNA polymerase inserts a dideoxynucleotide instead of the normal deoxynucleotide. The fragments are size separated on ABI sequencers with a maximum resolution of 1 bp. Each of the four dideoxynucleotides is labelled with a different fluorochrome tag which fluoresces at a characteristic wavelength when illuminated by a laser beam. Emitted fluorescence is collected by a camera and stored in a form of digitally encoded signals. Sequencing analysis software is used to visualise the sequence and allow base calling at each data point.

ddNTP	Fluorochrome	Colour (as seen in a chromatogram)
A	dR6G	Green
C	dROX	Blue
G	dR110	Black
T	dTAMRA	Red

Table No. 7: Fluorescently labelled ddNTPs used in sequencing reaction and their corresponding colours.

Sequencing reactions using the Big Dye Terminator (v.2.0) Cycle Sequencing kit (Applied Biosystems) were performed on PCR products that had previously been cleaned from unincorporated primers and dNTPs. For each fragment two reactions

were performed: with sense and anti-sense PCR primer. Reaction mix details are presented in Table No. 8, and thermal cycling conditions in Table No. 9.

Table No. 8: Sequencing reaction mixture preparation. All values apply for ONE reaction.

Mixture component	Volume (μ l)
Cleaned PCR product	5.00
Big Dye reaction mix	4.00
Primer (10 pmol/ μ l)	0.35
H ₂ O	0.65

Table No. 9: Sequencing reaction thermal cycling protocol.

Cycling Step	T°	Time	No of cycles
Denaturation	96°	2 min	1
Denaturation	96°	30 sec	25
Annealing	55°	15 sec	
Extension	60°	4 min	

Before sequencing reactions could be loaded onto a sequencer they had to be purified from any unincorporated nucleotides, salts and other impurities. That was done by gel filtering of sequencing products using MultiScreen HV filtration plates (Millipore) loaded with Sephadex G-50. The filtration plate had to be prepared few hours in advance (330 μ l ddH₂O added to each of the wells containing dry Sephadex) to allow the resin to swell at an ambient temperature. Excess water was spun out of the filtration plates in a centrifuge at 2500 rpm for 5 min, after which individual sequencing products could be added to the centre of individual columns. The filtration plate was then placed on top of a clean 96-well collection plate and spun at 2500 rpm for 5 min. Eluted products were dried in a speed vacuum at 65°C for 45min.

Final step involved re-suspension of dried sequencing products in 10 μ l HiDi formamide (Applied Biosystems) and their electrophoresis on ABI Prism 3100 Genetic Analyzer using a 36cm capillary array and POP-6 polymer.

DNA sequence was determined and polymorphisms characterised using a data analysis software Sequencher.

3.6 Genotyping

3.6.1 Restriction Fragment Length Polymorphism (RFLP)

Restriction fragment length analyses uses restriction endonuclease (RE) to digest DNA at specific 4-6 bp recognition sites. Sample DNA is cut (digested) with one or more REs and resulting fragments are separated according to molecular size using gel electrophoresis. Molecular size standards (*i.e.* 1kb ladder) are used to estimate fragment sizes of digestion products. Ethidium bromide (intercalating dye that fluoresces under UV light) is used for staining and revealing the fragments under UV (260 nm) light. Polymorphisms like base substitutions, additions and deletions result in sequence rearrangements within RE recognition sequences and in a subsequent loss/gain of a restriction enzyme cut site. This, in turn, results in digestion products of different sizes that are readily observed on gel electrophoresis.

Details on RFLP genotyping assays used in the study are described in Chapter 5.

3.6.2 Genotyping with microsatellites

Microsatellite DNA is defined as a small array of tandem repeats that are simple in sequence and occur regularly throughout the genome (approximately every 6 kb) ¹⁸⁷.

The high heterozygosity of microsatellite markers makes them suitable for population based association studies.

Many earlier studies used radioactive labelling of the PCR primers for microsatellite genotyping. The PCR products were run on a long polyacrylamide gel and alleles visualised by autoradiography. Between three to five markers of non-overlapping size could be run simultaneously. The method was effective for small studies, but it was time consuming and allele calling was often problematic.

Today, a fluorescent-based high-throughput technology provides a more efficient method for microsatellite genotyping. It can successfully be applied to both individual and pooled DNA samples. Up to six PCR product of non-overlapping size can be run concurrently. In addition, as one primer of each pair is labelled with one of the three available fluorescent labels, even products of overlapping size can be run together (see Figure No. 18).

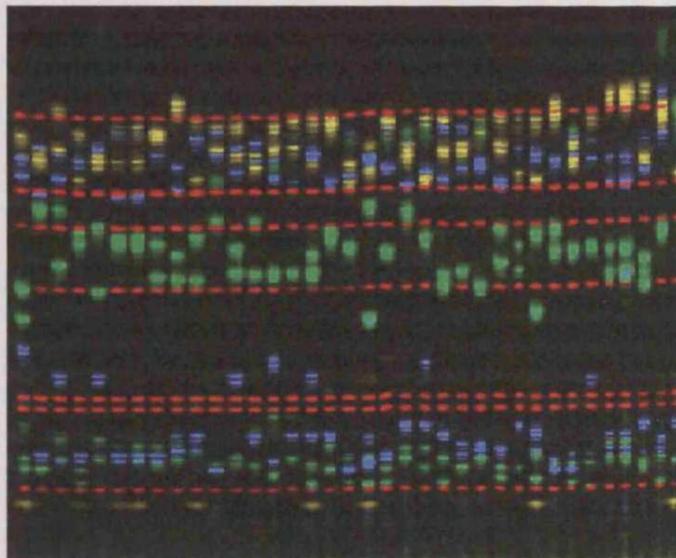


Figure No. 18: Image of a 96-well gel generated using GeneScan software. The size standard (TAMRA) appears red, the fluorescent microsatellite labels FAM, TET and HEX appear blue, green and yellow, respectively (with a kind permission of Dr N. Williams, Psychological Medicine, UHW, Cardiff)

More details on selection of microsatellite markers and genotyping using fluorescently labelled primers are given in Chapter 4 (individual and pooled DNA samples).

3.6.3 The SNaPshot™ Multiplex System assay genotyping

The chemistry is based on the dideoxy single nucleotide extension of an unlabeled oligonucleotide primer. The primer binds to a complementary template in the presence of fluorescently labelled ddNTPs and AmpliTaq DNA Polymerase. The polymerase extends the primer one nucleotide, adding a single fluorophore-labeled ddNTP to its 3' end. Because the primer is designed to anneal directly adjacent to the variant base of interest and the reaction does not include dNTPs, incorporation occurs only at a single site. Each of the four possible dye labelled terminators in a SNaPshot ddNTP Primer Extension reaction is labelled with a different fluorescent dye. The labelled primer extension products are detected and analyzed by the ABI 3100 DNA Analyzer.

ddNTP	Dye	Label Color of Analyzed Data
A	dR6G	Green
C	dTAMRA	Black
G	dR110	Blue
T (U)	dROX	Red

Table No. 10: Individual ddNTPs and their assigned fluorescent dyes.

A disadvantage of the method is that it is prone to yielding false positives. When running SNaPshot, negative control reactions always must be included, and if they do

not give the expected negative result, the rest of the data should be considered flawed. If even a small proportion of primers falsely extend on a "hairpin" or duplex the result will be a false positive. Extension primers with a complex secondary structure may also result in a SNaPshot assay failure. To reduce a probability of a primer forming a "hairpin" a software *FP Primer 1.0.1b* - by Dobril Ivanov (http://m034.pc.uwcm.ac.uk/FP_Primer.html) was used to assess sequence flanking a SNP in question. A flanking stretch of nucleotides with the least amount of a secondary structure would be selected and used as an extension primer. Detailed list of extension primers used in the CaMKII- α candidate gene study for SNaPshot genotyping is presented in Chapter 6.

Stage	Protocol
PCR reaction	See above
Post-PCR cleaning with SAP and <i>ExoI</i>	See above
SNaPshot reaction	See below
Post-SNaPshot cleaning with SAP	2.0 μ l H ₂ O and 2 μ l SAP (1U/ μ l) are added to 10 μ l SNaPshot reaction product and incubated at 37°C for 75 min
Preparation for analysis on ABI sequencers	2.0 μ l cleaned SNaPshot product with 8 μ l HiDi Formamide
Store frozen (-20°C)	

Table No. 11: Essential procedures involved in SNaPshot genotyping. Some of the stages were described in more detail earlier in the Chapter.

Table No. 12: SNaPshot reaction preparation

SNaPshot reaction component	Volume (μ l)
PCR product (cleaned with <i>ExoI</i> and SAP)	2.0
H ₂ O	2.0
Reaction buffer	2.5
SNaPshot Ready Reaction Mix	2.5
Extension primer (0.5 pmol/ μ l)	1.0

Table No. 13: SNaPshot reaction thermal cycling protocol

Cycling Step	T°	Time	No of cycles
Denaturation	94°	2 min	1
Denaturation	94°	5 sec	25
Annealing	43°	5 sec	
Extension	60°	5 sec	

The SNaPshot system can successfully be used for individual and pooled DNA genotyping. However, great attention has to be paid to interpreting pooled samples genotyping results as primer extension products are not equally represented for all alleles of a given genetic marker. Differential PCR amplification of alleles ¹⁸⁸ and differential efficiency in ddNTP incorporation ¹⁸⁹ are well known reasons for unequal representation of alleles. Fluorescent dyes used in the assay have unequal emission energies and result in different peak heights even when an individual heterozygous DNA sample is assayed. Therefore, estimated pool allele frequencies have to be corrected.

In order to allow for unequal allele representation, a simple and universal method ²⁰ can be used to correct estimated allele frequencies from pools (see Fig. No. 19).

$$A_{\text{corrected}} = A / (A + kB)$$

Figure No. 19: Correction of estimated allele frequencies from pools. Frequency of allele A is calculated by applying the above formula, where A and B are the peak heights of the primer extension products (representing alleles A and B in pools) and *k* is the mean of replicates of A/B ratios observed in a heterozygote ²⁰.

3.7 Statistical Analysis

Statistical tools employed for analysis of data collected from different studies are described in relevant chapters.

4 GENOME-WIDE SCREEN FOR ASSOCIATION WITH g

4.1 Introduction

4.1.1 Background of the study

General cognitive ability, g , is one of the most heritable human behavioural traits³³. A model-fitting analysis of cumulative data from more than a hundred world-wide studies on intelligence⁸⁷ suggests that about half of the variance in IQ scores can be attributed to genetic factors, most of which are additive in nature³³. The next natural step in studying general cognitive ability would be to identify some of the genes responsible for its high heritability.

The complex nature of inheritance and a roughly normally distributed IQ scores suggest that universally observed variation in intelligence is governed by a mixture of genetic and environmental factors. The exact number of genes, quantitative trait loci, involved in shaping g is not known, and even the most accurate estimate is likely to be conservative (reasons for this are discussed in paragraph 1.4.2 of the General Introduction). The total genetic effect on the trait is not evenly distributed between genetic loci. The most plausible scenario will probably involve few modest effect loci and a large number of small effect loci^{63, 67}.

A small effect QTL is usually described as a locus that accounts for only about 1% or less of the variance¹⁶². Finding genes of such a small effect requires the power of an association rather than a linkage study design¹⁰⁹. In the first attempt to identify some of the genetic loci contributing to variance in g , Plomin *et al.* (1994) genotyped

100 DNA markers in or near genes involved in brain functioning, but no replicated associations with *g* were found ¹⁶³.

Instead of a candidate gene approach, an allelic association study can be made more systematic if a large number of densely distributed markers are genotyped along a chromosome. Chorney *et al.* (1998) made a first step in that direction by typing 47 microsatellite markers on the long arm of chromosome 6 ¹⁶⁴. A significant association for a marker positioned within the gene for insulin-like growth factor-2 receptor (IGF2R) was successfully replicated by the same group using larger samples ¹⁶⁵. Considering how difficult it is to find a true association, this was an encouraging result, which offered justification for further development of the strategy and continuation of the genome screen in search of association with *g*.

4.1.2 DNA pooling

4.1.2.1 DNA pooling makes large genome-wide scans feasible

The problem with using a dense map of markers for whole-genome screens is the amount of genotyping required. Linkage disequilibrium with microsatellite markers is typically found within a distance of 500 kb (~ 0.5 cM) ¹⁹⁰. To cover the human genome with markers spaced at no more than 1 cM, a density desirable for a systematic genome screen for allelic association studies, one would need to genotype about 3500 microsatellite markers. This would ensure that, on average, any trait locus is less than 0.5 cM (~ 500 kb) from the marker. For LD studies using bi-allelic markers the situation is much worse since the suggested number of markers varies between 35,000 ¹³², >100,000 ¹³³, 250,000 ¹⁴⁶ and most pessimistic 1,000,000 ¹¹⁷. For power to be large enough to detect small effect loci, one also has to aim for larger sample sizes (several

hundred). This would require the creation of several million genotypes, which is beyond the capabilities of most research laboratories.

However, the problem can be overcome by adopting a DNA pooling strategy^{191, 192}. The need for genotyping is reduced by pooling DNA from all individuals within a group (both case group and control group) and then genotyping the pooled groups. So, for example, instead of genotyping 202 individuals in the first stage of the IQ genome scan (101 high *g* and 101 average *g* controls), we need to generate only six pool amplifications per microsatellite marker if the pools are amplified in triplicates (for description, see Chapter 2: General Materials and Methods). This method allows a rapid screening of a large number of markers and an efficient identification of those that are most likely to show allelic association. The method has been applied successfully in QTL association studies of both humans and animals^{193, 194}.

The strategy also benefits from the possibility of selective pooling of individuals. It has been shown that greater power can be achieved by selecting individuals at the phenotypic extremes of the population¹⁹⁵, and that fewer individuals need to be genotyped in order to retain the same level of power as with an unselected sample¹⁹⁴. In comparison with standard selective genotyping, the selective pooling means doing even fewer genotypings per marker while only slightly decreasing the power (assuming the allele frequencies are estimated without error)¹⁹⁵.

It is worth noting that the collection of sufficiently large phenotypically extreme samples would require that initially a relatively large number of individuals be phenotyped. Nevertheless, the benefits associated with selective pooling justify greater initial efforts in phenotyping. With quantitative traits such as *g*, different lower thresholds can be chosen for cases (and controls). Table No. 14 gives examples of sample sizes that would be required for the IQ QTL study if different lower thresholds were to be used for selecting individuals for a high *g* group while maintaining the same level of power (in this example set to 95%).

Total QTL variance	Distance of the selected High <i>g</i> group from the mean			
	+1 SD	+1.5 SD	+2 SD	+ 2.5 SD
0.05	194	118	78	54
0.025	404	246	162	113
0.01	1047	641	423	295

Table No. 14:

Estimates of sample size necessary for detection of QTLs that account for 5%, 2.5% and 1% of the total variance, assuming complete linkage disequilibrium ($D'=1$). For the original set, 101 cases (high *g* individuals) and 101 controls (average *g* individuals) were considered. Lower threshold for cases was taken to vary between 1 SD and 2.5 SD above the mean. Controls were selected to be ± 0.67 SD from the mean; QTL and marker allele frequencies were set to 0.25.

4.1.2.2 Problems associated with DNA pooling

Unlike individual genotyping, DNA pooling cannot provide us with real allele frequencies for a marker typed on pools. To be able to analyse data from pooled DNA amplifications and successfully employ them in allelic association studies, we must estimate marker allele frequencies.

The power of selective DNA pooling, as strong as it could be for detecting genes with small effects, can decrease steeply with the increase of error in estimating allele frequencies¹⁹⁵. The errors mainly stem from three PCR artefacts: stutter and differential amplifications, both commonly associated with microsatellite repeats, and non-templated nucleotide addition.

Stutter bands result from the amplification of products one or two, and rarely more, repeat units shorter than the correct sized amplicon. This is thought to be due to slippage of *Taq* DNA polymerase on the repeated sequence¹⁹⁶, and/or out-of-register

annealing of truncated PCR products ¹⁹⁷. Stutter is marker specific and occurs more frequently in di-nucleotide repeats, although it can be seen, to a lesser extent, in tri- and tetra-nucleotides. The stutter bands overlap and add to the total signal coming from real alleles, making it difficult for us to determine accurately the frequencies of pooled alleles.

Differential amplification is a PCR artefact observed in heterozygotes, and it is a consequence of preferential amplification of the smaller allele. In some extreme circumstances, differential amplification may result in complete disappearance of the larger allelic product ¹⁹⁸. This is thought to result from longer alleles re-annealing at a faster rate, which reduces the efficiency of PCR amplification ¹⁹⁸.

Non-templated addition of a nucleotide to the 3' end of amplification products, catalysed by *Taq* DNA polymerase, is a marker-specific activity and presents another potential source of error in genotyping studies ¹⁹⁹. Which nucleotide will be added depends on the type of polymerase used as well as the identity of the last residue of the DNA template ²⁰⁰. Various DNA polymerases can add nucleotides with varying efficiencies ²⁰⁰. Smith *et al.* estimate that between 1% and 3% genotyping error rate is attributable to this phenomenon ²⁰¹.

4.1.2.3 Solving problems

Several groups have attempted to solve some of the problems associated with previously described PCR artefacts. Two main approaches can be recognised: practical and mathematical.

I) **Practical solutions:** Practical solutions have been offered for planning stages, when a researcher is advised to choose for his/her study only stutter-free microsatellite

markers. These can be tetra-nucleotide²⁰² and/or carefully selected di-nucleotide repeat markers²⁰³. However, solving the stutter problem by using tetra-nucleotides may result in an exaggerated problem of differential amplification²⁰².

It has been suggested that the problem of differential amplification can be avoided by raising the molar ratio of the *Taq* DNA polymerase to the template DNA during the PCR²⁰⁴, or simply by decreasing the amount of DNA template subjected to the PCR²⁰².

The tendency to preferentially amplify lower molecular weight alleles can also be reduced by partial replacement of the physiologic dGTP by 7-deaza-2'-dGTP during amplification, which weakens the stability of inter-molecular G-C base pairing and ensures that DNA polymerase successfully completes primer extension^{205, 206}.

Demers *et al.* (1995) suggested addition of peptide nucleic acid (PNA) to PCR reaction. PNA is a DNA mimic with a unique ability to hybridise to DNA template and make it unavailable for premature re-annealing with a complementary DNA strand while allowing polymerase to displace it and extend the primer to completion¹⁹⁸.

There are also a few options for treatment of non-templated addition of a nucleotide at the 3' end: **1)** employment of a high fidelity (*Pyrococcus furiosus*) *Pfu* DNA polymerase, which is highly efficient in making blunt-ended PCR products²⁰⁰, **2)** "pig-tail" primers, where a non-labelled primer of each pair has a consensus sequence tail (GTTTCTT) appended to its 5' end and produces nearly 100% adenylation of the 3' end of the forward strand²⁰⁷, **3)** post-PCR treatment of amplified products with *E. coli* DNA polymerase I Klenow fragment to repair 3' overhanging ends and generate blunt-ended products, **4)** T4 DNA polymerase treatment of PCR products²⁰⁸, or **5)** prolongation of the final extension step at 72°C following the amplification to 90 min, the method being probably most amenable to automated high throughput genotyping²⁰¹.

II) **Mathematical solutions:** Le Duc *et al.* (1995) improved the accuracy in the estimation of allele frequencies by arithmetic removal of stutter bands²⁰⁹. They measured peak heights of all alleles, but before converting them into allele frequencies the stutter ratio was applied as a correction factor. The method was tested on two

microsatellites, D11S903 and D14S80, and estimated allele frequencies reflected real allele frequencies quite accurately ²⁰⁹. However, the stutter ratio was calculated on the basis of the amount of stutter from the smallest allele only and not adjusted for the fact that the amount of stutter is allele-dependent.

More accurate mathematical stutter-correction was developed by Perlin *et al.* (1995). This method is based on the finding that each microsatellite marker has its own very reproducible stutter pattern. Also, rather than assuming an invariant stutter pattern for all alleles, Perlin *et al.* advise employment of algorithms that permit allele-dependent variation. Such algorithms showed, on average, less error by a factor of 10 in estimating allele frequencies than those assuming a constant stutter pattern ²¹⁰.

Barcellos *et al.* (1997) have combined Perlin's method with mathematical correction for differential amplification. A preferential-amplification factor is calculated as a ratio of peak heights (X/Y), where X denotes the smaller peak. In the pooled samples, in addition to Perlin's correction for stutter, observed values for peak heights are further modified and divided by the preferential-amplification factor prior to conversion to allele frequencies ¹⁹¹.

Even though Perlin and Barcelloes created methods capable of producing quite accurate allele frequency estimates from pooled data, each individual marker required a lengthy work-up, which would be a substantial slowing-down factor in large-scale studies requiring many markers to be tested.

4.1.2.4 The Δ AIP method

Dealing with differential amplification and stutter problems, at either practical or mathematical level, can be very challenging for studies planning to test thousands of microsatellite markers. To by-pass these problems altogether, Daniels *et al.* (1998)

developed a simple method for analysis of pooled genotypes that relies on statistical comparison of Allele Image Patterns (AIPs) for pooled cases and pooled controls ¹⁹². For more detailed description of the method see Materials and Methods section.

The main difference between this and other methods is that the Δ AIP (Allele Image Pattern Difference) method is used as a rapid initial screening step to identify markers for individual genotyping. Its goal is not to estimate absolute allele frequencies for a single group (cases or controls) which requires adjustments for technical problems discussed, but simply to compare the uncorrected products of pooled microsatellite amplification.

The method does not require lengthy analysis of each marker prior to genotyping, which allows the application of DNA pooling to large-scale association studies.

4.2 Aim

The aim of this study was to perform a genome-wide screen for the association with general cognitive ability, g , using a large number of densely (1 cM grid) spaced microsatellite markers. To make the study feasible, completed in time and within available resources, selective DNA pooling and the Δ AIP method were employed. A multi-stage study design was used to obtain a balance between false positives and false negatives in the search for QTLs of small effect size.

4.3 Materials and Methods

4.3.1 Sample

See General Materials and Methods

4.3.2 Microsatellite marker selection

This computer-intensive initial stage of the study was carried out mainly by the project's post-doctoral research fellow, Paul J. Fisher, with some input from Linzy Hill (collaborating group in London IOP) and myself.

DNA markers were selected from the Genetic Location Database (LDB) summary maps (http://cedar.genetics.soton.ac.uk/public_html/summaryml.html) and ordered according to sex-averaged genetic distances available at the Marshfield Centre for Medical Genetics web site (http://research.marshfieldclinic.org/genetics/Map_Markers/maps/IndexMapFrames.html). When no genetic distance information was available, markers were positioned according to the physical map. Ideally, markers were 1 cM apart and had between 5 and 9 alleles and high heterozygosity scores (0.5 and 0.9). However, due to scarcity of markers in some regions, some less than ideal markers were chosen.

In total, 2871 microsatellite markers were selected for a screening set using the above conditions. Primer sets with fluorescently labelled reverse primers were purchased from MWG-Biotech in Germany. Additional 117 primer pairs (ABI PRISM® Linkage Mapping Set v2.5) were used from marker stocks already available at the laboratory. This brought the total marker number to 2988 with average inter-marker genetic distance of 1.5 cM, and with ~80% of the genome within 1 cM of a marker.

Coverage success varied between and within chromosomes, with somewhat poorer coverage toward telomeric regions.

Details on the 1847 microsatellite markers (allele numbers, heterozygosity scores, amplicon sizes, primer sequences) that were successfully genotyped on pools using standard PCR conditions (described in section 4.3) can be found on the web site developed by the London project team (<http://www.sgdp.org.uk>).

4.3.3 Δ AIP statistic calculation

Allele image patterns are produced for cases and controls by a standard ABI Genescan software, overlaid in Genotyper (see Figure No. 20) and imported into commercially available graphics software Debabelizer (Equilibrium).

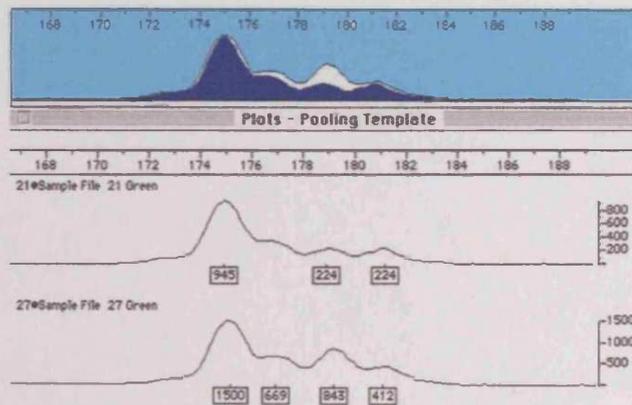


Figure No. 20:

Overlaid Allele Image Patterns (AIPs) with differently coloured shared (dark blue) and non-shared (white) areas; area sizes are determined from the pixel number for each of the colours representing shared and non-shared areas.

The ΔAIP statistic is calculated as a function of the total shared and non-shared area, whose sizes are determined from the pixel number of colours used to distinguish the areas:

$$\Delta AIP = \frac{\text{Non-shared area}}{\text{Non-shared} + \text{Shared area}}$$

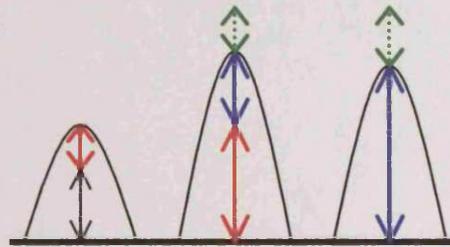
The statistic is an estimate of relative difference in allele frequencies between the groups. The significance of the ΔAIP is then determined by a computer simulation program ¹⁹².

4.3.4 Allele Specific Test (AST) statistic calculation

The test incorporates a simple algorithm for "blanket" correction of stutter and differential amplification. Algorithms for di-, tri- and tetra-nucleotide repeat markers were derived empirically by Paul J. Fisher (unpublished). Pearson χ^2 with a 2x2 contingency table is used to compare estimated frequency of the target allele versus all other alleles.

Figure No. 21:

Simplified explanation of the algorithm used for stutter and differential amplification correction prior to AST analysis. Three peaks from amplified pooled DNA are shown. Differently coloured arrows denote various sources contributing to peak height generation. So, for example, middle peak height will have to be reduced for the amount of the stutter coming from the right peak (blue arrow), increased for the amount of its own stutter present in the left peak (red arrow) and finally increased for the amount of differential amplification expected for the particular allele (green dashed arrow).



4.3.5 Multi-stage study design

As previously suggested by Risch and Teng (1998), studies using pooled DNA should ideally have at least two stages: first, screening DNA pools by pooled genotyping and second, individual genotyping of positive markers identified in the initial screen²¹¹. In this study, a four stage design with two independent samples was used in order to provide balance between false positive and false negative results in search for QTLs of small effect size (See Figure No. 22).

Stage 1: Original case/control sample – DNA pools

This is the initial screening stage in which all markers were tested on pooled samples of original 101 high *g* individuals and 101 average *g* controls. Δ AIP and simulated *p*-values are calculated for each of the successfully amplified microsatellites.

Markers yielding $p < 0.05$ were subjected to an Allele Specific Test (AST) in order to establish which alleles show the largest difference between the high g and control group. Only markers with $p < 0.05$ for both ΔAIP and AST qualified for the Stage 2.

Stage 2: Replication case/control sample – DNA pools

The allele-specific directional hypothesis from Stage 1 was tested in the replication case-control sample using pooled DNA. A one-tailed test of significance was used because the target allele was required to show allele frequency difference between the groups in the same direction as in Stage 1. For a marker to qualify for the next stage, a p-value had to be < 0.05 .

Stage 3: Individual genotyping of the original sample set

Markers surviving Stages 1 and 2 were individually genotyped on the original case/control sample. The allele-specific directional hypothesis was tested for significance using one-tailed Pearson χ^2 test comparing that allele against all others.

Stage 4: Individual genotyping of the replication sample set

Finally, Stage 3 markers with $p < 0.05$ were individually genotyped on the replication case/control set. Pearson χ^2 test (one-tailed) was used to assess significance of allele-specific directional hypothesis.

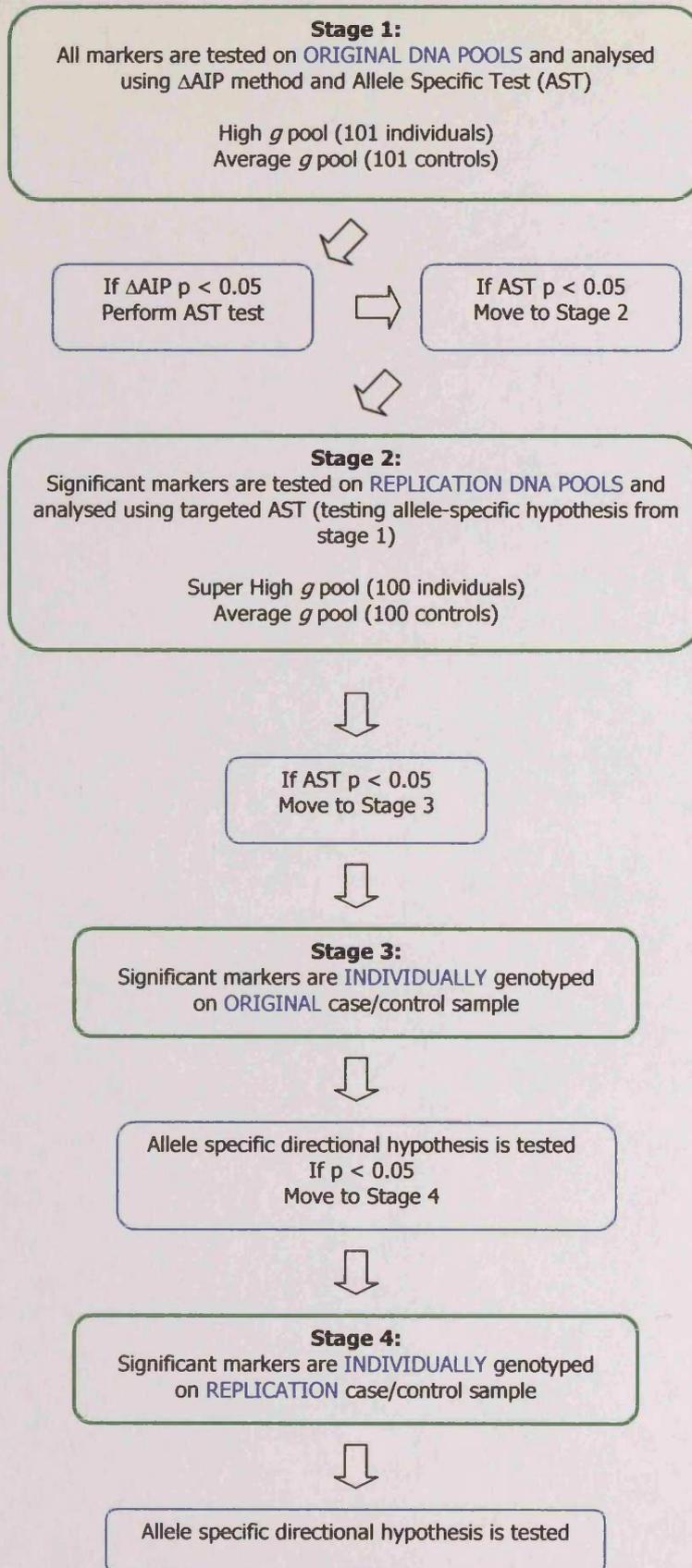


Figure No. 22: Graphical presentation of a multi-stage study design adopted for the genome-wide screen for association with general cognitive ability.

4.3.6 DNA pooling

See General Materials and Methods

4.3.7 Pool genotyping

Each of the three replica pools for "high" *g* and "average" *g* was amplified at the same time, using a touchdown PCR protocol. For each microsatellite marker a separate reaction was performed. Amplification products were pooled while care was taken to combine different fluorescent dyes (Hex, Fam and Tet) and product sizes. Up to four different markers were loaded into one gel lane. Because of the difference in fluorescing intensity, PCR products had to be diluted prior to pooling: Fam labelled 8x, Tet 6x and Hex 5x. Diluted and pooled PCR products (1.5µl) were mixed with loading dye (1.5µl) and GeneScan®-500 size ladder (0.5µl) and size separated on 12.5% denaturing gels using ABI373A sequencer.

Gels were analysed and allele image patterns (AIPs) generated for each of the products with the GENESCAN® software. Visualisation and overlaying of AIPs was carried out using GENOTYPER®1.1.1. Only one of the three AIP images generated by each of the two groups was selected as a consensus image.

In order to compare the results of the pooled genotyping between high *g* and control group, consensus images were overlaid and imported into a graphics software DeBabelizer™. Different colours were used to fill in shared and non-shared areas (see paragraph No. 4.1.2.4), whose sizes were then expressed by the software as a number of pixels and used in the formula for Δ AIP calculation. Markers were tested for significant Δ AIP using a computer simulation program (for description see paragraph No. 4.3.3).

Allele counts and frequencies were estimated from pooled data using the algorithm described in section No. 4.3.4. Target allele significance was tested using Pearson χ^2 .

4.3.8 Simulated p-value

Absolute ΔAIP value does not realistically reflect the significance of difference in allele frequencies observed between two pools because the significance of ΔAIP will depend on the number of parameters, such as sample size, number of marker alleles and allele frequency. The test for significance was performed by a statistical method based on a computer program, which, with 2000 simulations per test, simulates case and control samples from a population with allele frequencies estimated from the peak heights of the control sample ¹⁹².

4.3.9 Individual genotyping

For individual genotyping, PCRs and gel runs were performed using the same protocol, equipment and software as described for pooled genotyping, except for the last step, which involves image overlaying and ΔAIP calculation. The significance of the target allele was estimated using Pearson χ^2 .

4.3.10 Power

Power analysis was carried out using publicly available software "Genetic Power Calculator" (http://statgen.iop.kcl.ac.uk/gpc/index.html#qcc_ins). All calculations in the program are based upon formula derived by Sham *et al.* (2000)²¹². Detailed instructions on entering values into required fields are available on the web site.

Calculations were performed separately for original and replication sample sets. Nine different scenarios were considered: QTL that accounts for 5%, 2.5% and 1% of the phenotypic variance, assuming three different levels of LD ($D'=1$, 0.75 and 0.5). The results are presented in Tables No. 15 and No. 16.

Total QTL variance	Power to detect QTL ($\alpha=0.05$)		
	$D'=1$	$D'=0.75$	$D'=0.50$
0.05	100%	96%	73%
0.025	93%	74%	42%
0.01	56%	36%	19%

Table No. 15: Power estimates used to detect QTLs that account for 5%, 2.5% and 1% of the total variance under different levels of linkage disequilibrium (D'); For the original set, 101 cases (high g individuals) and 101 controls (average g individuals) were considered; Lower threshold for cases was taken to be 2.5 SD above the mean; Controls were selected to be ± 0.67 SD from the mean; QTL and marker allele frequencies were set to 0.25.

Total QTL variance	Power to detect QTL ($\alpha=0.05$)		
	$D'=1$	$D'=0.75$	$D'=0.50$
0.05	100%	100%	99%
0.025	100%	99%	91%
0.01	97%	85%	53%

Table No. 16: Power estimates used to detect QTLs that account for 5%, 2.5% and 1% of the total variance under different levels of linkage disequilibrium (D'); For the replication set 100 cases (super-high g individuals) and 96 controls (average g individuals) were considered; Lower threshold for cases was taken to be 5 SD above the mean; Controls were selected to be ± 0.67 SD from the mean; QTL and marker allele frequencies were set to 0.25.

4.4 Results

4.4.1 General notes

Genome-wide screen workload was shared between Cardiff and London groups collaborating on the IQ QTL project. All DNA pools were created in Cardiff: the original by myself, and replication by the project's post-doctoral research fellow Dr Paul Fisher. Both groups were assigned a set of chromosomes to work on. Each of the chromosomes was expected to be carried through all the stages by the same group it initially had been assigned to. The exception were the ABI linkage panel microsatellite markers, which were tested in Cardiff regardless of their chromosomal position. I was responsible for genotyping most of the microsatellite markers assigned to Cardiff group (~1600 markers). All Δ AIPs, a proportion of the simulated p-value calculations and some AST analysis for Cardiff markers were calculated by myself, while the rest of the simulated p values, AST and other statistical tests were carried out by Dr Paul Fisher. The analysis of markers tested in London was carried out by the London project group.

Chr. No.	Average inter-marker distance (cM)	Largest inter-marker distance (cM)	Chr. No.	Average inter-marker distance (cM)	Largest inter-marker distance (cM)
1	2.18	10.01	12	1.65	5.76
2	1.77	9.04	13	1.54	7.82
3	1.56	5.59	14	2.38	8.58
4	2.01	13.55	15	2.66	11.63
5	1.83	6.47	16	2.05	10.33
6	1.78	9.18	17	2.68	9.60
7	1.78	8.94	18	2.91	19.37
8	2.14	7.52	19	2.87	10.97
9	2.31	11.70	20	2.35	8.72
10	2.86	10.32	21	2.23	8.05
11	2.71	9.59	22	2.57	10.38

Table No. 17: Average inter-marker distances (cM) for each of the chromosomes analysed. Limited microsatellite availability resulted in a number of gaps in our final map. Largest gaps (cM) for each of the chromosomes are presented.

4.4.2 Stage 1

Of the 2988 microsatellite markers tested, Δ AIP and simulated p-values were obtained for 1847 (62%). Using standard touchdown PCR conditions (see General Material and Methods), 1588 markers (53%) yielded amplification products with highly replicable allele image patterns, where at least two out of three AIPs generated for each of the groups gave nearly complete overlays. The remaining 1400 markers went through a second amplification protocol, using *Taq* Gold polymerase (see General Materials and Methods). This generated replicable AIPs for another 259 markers (9%).

	Total No. of markers tested	No. of ΔAIP_{ori} results (%)	No. of markers with simulated p_{ori}<0.05 (%)	No. of markers with AST p<0.05 (%)
Cardiff	1641	1021 (62%)	51 (5%)	38 (75%)
London	1347	826 (61%)	186 (23%)	70 (38%)
TOTAL	2988	1847 (62%)	237 (13%)	108 (46%)

Table No. 18: Total number of markers considered for the IQ QTL study and their progress through different levels of the Stage 1 are presented for each of the two collaborative centres (Cardiff and London). Cardiff markers come from chromosomes 2, 3, 4, 5, 6, 7, 12, 14, 15, and 17. Cardiff centre tested only a very small proportion of microsatellite markers belonging to other chromosomes (those that were part of the ABI linkage mapping panel).

For a large number of markers (~1/3 of all markers purchased), Δ AIP could not be calculated for various reasons. First, for a significant proportion of the markers no amplification product could be recovered using standard PCR conditions with either of the *Taq* polymerases. Second, for a number of microsatellites amplicons were of poor

quality, which meant that images could not be overlaid and could not be used for Δ AIP analysis. Third, in some instances non-templated addition of a nucleotide at the 3' end was very pronounced and could not be resolved by Klenow fragment treatment of the product. Some markers had to be discarded because AIPs showed significant variation within the group and no consensus image could be selected as a representative of that group.

Out of 1847 generated Δ AIPs, 237 had significant simulated $p < 0.05$. Only 92 markers were expected to be significant by chance alone with $p < 0.05$. More than three quarters of positive markers came from the London group (see Table 18). The most likely explanation was generation of false positive results through accepting less than perfect AIP overlays. A large proportion of these (62%) were filtered out after the AST analysis was performed. In contrast, only 5% of Cardiff markers had significant p -values before the AST analysis, with most of them (75%) staying significant after further analysis, possibly due to more rigorous criteria for accepting AIP overlays.

Of the total of 237 markers, 108 had significant AST results ($p < 0.05$) and qualified for the second stage.

4.4.3 Stage 2

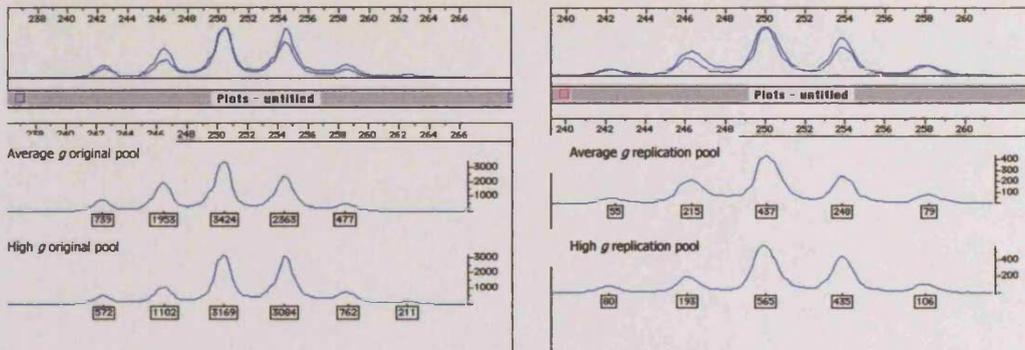
108 markers reached Stage 2, where they were tested on independent replication sets of pooled DNA samples (Stage 2 markers that were done in our laboratories are listed in Table No. 19). Of these, 7 (D2S427, D3S2425, D4S2460, D4S175, D6S277, D16S687 and D18S1110) showed significant simulated p -values ($p < 0.05$) when their AIP images were analysed, while only 5 were expected to be significant by chance alone. Four markers (D3S2425, D4S175, D16S687 and D18S1110) did not have a significant AST result, or they had a significant AST, but their association was in the

opposite direction (*i.e.* allele was more frequent in high *g* individuals than in controls in the original sample set, but was significantly less frequent in high *g* individuals than in controls in the replication set, and *vice versa*; which was observed for D3S2425), or they had a different allele showing association in the replication set (observed for D6S277).

No.	Marker ID	Simulated P _{ori}	AST P _{ori}	No.	Marker ID	Simulated P _{ori}	AST P _{ori}
1	D2S285	0.023	0.019	20	D5S1726	0.040	0.010
2	D2S2264	0.000	0.000	21	D6S277	0.020	0.039
3	D2S2277	0.004	0.038	22	D6S506	0.020	0.010
4	D2S1776	0.013	0.019	23	D6S1597	0.010	0.000
5	D2S2978	0.021	0.049	24	D6S1564	0.010	0.038
6	D2S427	0.026	0.007	25	D6S446	0.000	0.008
7	D3S1217	0.003	0.011	26	D7S1508	0.001	0.000
8	D3S2433	0.000	0.003	27	D7S1874	0.003	0.011
9	D3S2425	0.020	0.001	28	D7S2501	0.010	0.004
10	D3S3699	0.005	0.009	29	D12S1632	0.018	0.046
11	D3S3578	0.001	0.005	30	D12S90	0.003	0.005
12	MSX1	0.002	0.005	31	D12S1727	0.010	0.000
13	D4S404	0.046	0.015	32	D14S749	0.002	0.000
14	D4S3001	0.020	0.001	33	D14S65	0.040	0.028
15	D4S2460	0.009	0.027	34	D17S831	0.049	0.013
16	D4S175	0.026	0.004	35	D17S1863	0.004	0.000
17	D4S1607	0.019	0.003	36	D17S1794	0.009	0.048
18	D4S2943	0.038	0.015	37	D17S1299	0.023	0.021
19	D5S1980	0.000	0.000	38	D17S1350	0.040	0.000

Table No. 19: The microsatellite markers tested in Cardiff as part of Stage 2. Those typed in **blue** and bold had simulated $p < 0.05$ in the replication sample set.

Figure No. 23: D2S427 allele image patterns



Left
D2S427 allele image patterns generated by the ORIGINAL high g and average g DNA pools and the overlaid image used for Δ AIP calculation.

Δ AIP=0.16

sim p=0.03

AST p=0.01 (peak 2)

Right
D2S427 allele image patterns generated by the REPLICATION high g and average g DNA pools and the overlaid image used for Δ AIP calculation.

Δ AIP=0.15

sim p=0.04

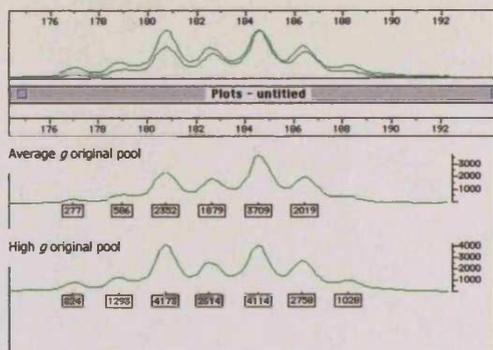
AST p=0.03 (peak 2)

D2S427	Description	Peak 1	Peak 2	Peak 3	Peak 4	Peak 5
Average g original pool	Uncorrected peak height	739	1953	3424	2363	477
	Corrected peak height	678	1869	3577	2997	572
	Estimated allele frequency (%)	7.0	19.3	36.9	30.9	5.9
High g original pool	Uncorrected peak height	572	1102	3169	3084	762
	Corrected peak height	550	968	3247	3479	914
	Estimated allele frequency (%)	6.0	10.6	35.5	38.0	10.0
Allele frequency difference (%)		1.0	8.7	1.4	-7.1	-4.1

D2S427	Description	Peak 1	Peak 2	Peak 3	Peak 4	Peak 5
Average g replication pool	Uncorrected peak height	55	215	437	248	79
	Corrected peak height	47	201	466	277	95
	Estimated allele frequency (%)	4.3	18.5	42.9	25.5	8.7
High g replication pool	Uncorrected peak height	80	193	565	435	106
	Corrected peak height	75	168	590	491	127
	Estimated allele frequency (%)	5.2	11.6	40.7	33.8	8.8
Allele frequency difference (%)		-0.8	6.9	2.2	-8.3	0.0

Tables No. 20 and 21: Calculation of estimated allele frequencies for D2S427 in original and replication DNA pools. Selection of a target allele (Peak 2) was based on the ORIGINAL pools' results. Target allele is printed in RED.

Figure No. 24: D4S2460 allele image patterns



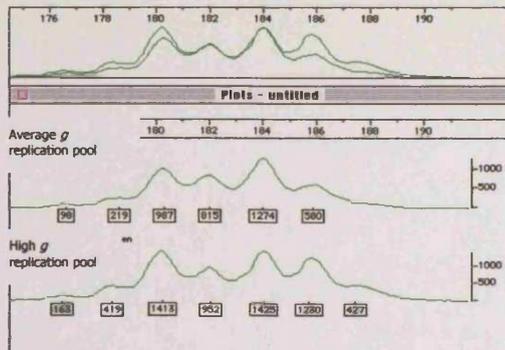
Left

D4S2460 allele image patterns generated by the ORIGINAL high g and average g DNA pools and the overlaid image used for Δ AIP calculation.

Δ AIP=0.22

sim p=0.01

AST p=0.03 (peak 5)



Right

D4S2460 allele image patterns generated by the REPLICATION high g and average g DNA pools and the overlaid image used for Δ AIP calculation.

Δ AIP=0.21

sim p=0.00

AST p=0.01 (peak 5)

D4S2460	Description	Peak 1	Peak 2	Peak 3	Peak 4	Peak 5	Peak 6
Average g original pool	Uncorrected peak height	277	586	2352	1879	3709	2019
	Corrected peak height	288	179	2554	1304	3965	2650
	Estimated allele frequency (%)	2.6	1.6	23.3	11.9	36.2	24.2
High g original pool	Uncorrected peak height	824	1293	4173	2614	4114	2758
	Corrected peak height	842	610	4567	2159	4198	3620
	Estimated allele frequency (%)	5.3	3.8	28.6	13.5	26.2	22.6
Allele frequency difference (%)		-2.6	-2.2	-5.2	-1.6	10.0	1.6

D4S2460	Description	Peak 1	Peak 2	Peak 3	Peak 4	Peak 5	Peak 6
Average g original pool	Uncorrected peak height	98	219	987	815	1274	580
	Corrected peak height	104	48	1049	648	1406	761
	Estimated allele frequency (%)	2.6	1.2	26.1	16.1	35.0	19.0
High g original pool	Uncorrected peak height	163	419	1413	952	1425	1230
	Corrected peak height	155	189	1523	834	1348	1614
	Estimated allele frequency (%)	2.7	3.3	26.9	14.7	23.8	28.5
Allele frequency difference (%)		-0.1	-2.1	-0.8	1.4	11.2	-9.6

Tables No. 22 and 23: Calculation of estimated allele frequencies for D4S2460 in original and replication DNA pools. Selection of a target allele (Peak 5) was based on the ORIGINAL pools' results. Target allele is printed in RED.

4.4.4 Stages 3 and 4

The two markers reaching Stage 3 (D2S427 and D4S2460) were individually genotyped on the original sample set (Table No. 24). The individual genotyping confirmed the results of DNA pooling in that both markers showed differences between the groups for the same allele and in the same direction as those detected in Stages 1 and 2 (D2S427, $\chi^2=2.76$, $p=0.048$ and D4S2460, $\chi^2=5.40$, $p=0.010$).

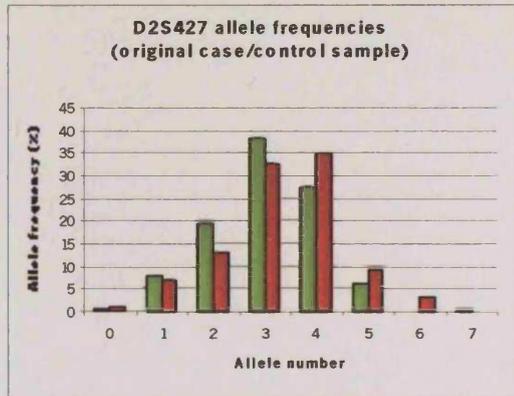
Allele specific results were also in the expected direction in the case-control replication sample. Stage 4 results were significant for one of the two markers, D4S2460 ($\chi^2=2.98$, $p=0.042$). Detailed genotyping results for the two markers are shown in Table No. 24. Graphical illustration of markers' allele frequencies can be seen in figures No. 25 and 26.

Table No. 24: Genotyping results for 7 markers qualifying for Stage 2.

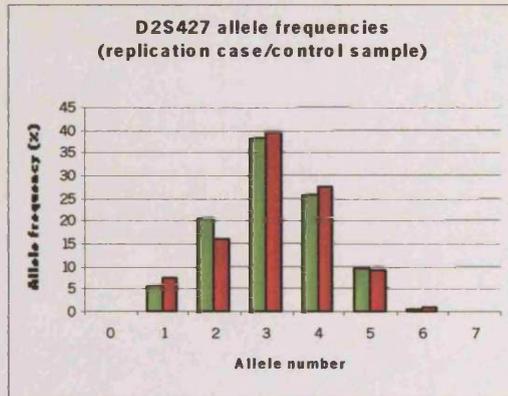
Marker (target allele)	Stage 1		Stage 2		Stage 3			Stage 4				
	Original case-control sample Δ AIP-sim p-value	AST p-value	Replication case-control sample Δ AIP-sim p-value	AST p-value	Original case-control sample High <i>g</i>	Control <i>g</i>	χ^2	p	Replication case-control sample High <i>g</i>	Control <i>g</i>	χ^2	p
D2S427 (2)	0.026	0.007	0.04	0.030	27 (13%)	39 (19%)	2.76	0.048	25 (16%)	38 (21%)	1.08	0.149
D3S2425 (4)	0.020	0.001	0.00	0.059 [†]								
D4S2460 (5)	0.009	0.027	0.00	0.013	56 (29%)	77 (40%)	5.40	0.010	50 (29%)	73 (38%)	2.98	0.042
D4S175	0.026	0.004	0.01	0.292								
D6S277 (3)	0.020	0.039	0.00	0.371								
D16S687 (4)	0.000	0.001	*	0.073								
D18S1110 (5)	0.000	0.035	*	0.160								

[†]Target allele in opposite direction to the original case-control sample set

*Data not available



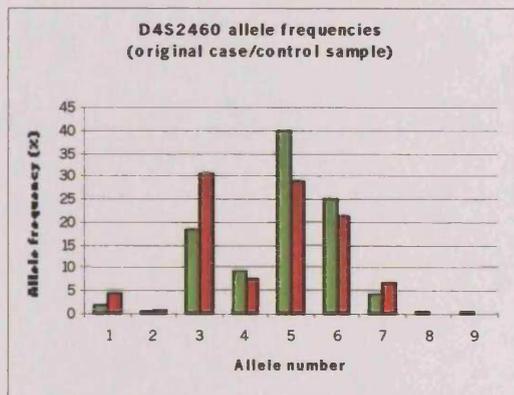
AST (allele 2): $\chi^2 = 3.13$
 $p = 0.038$



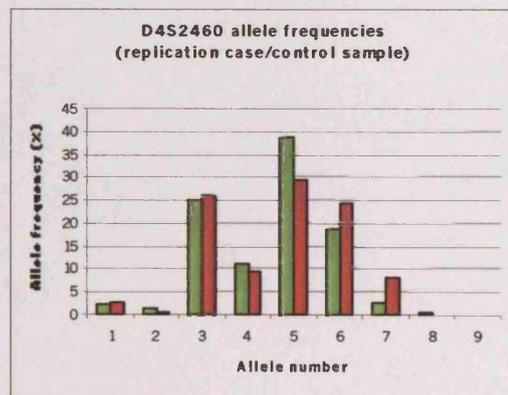
AST (allele 2): $\chi^2 = 1.25$
 $p = 0.132$

Average g individuals
 High g individuals

Figure No. 25: Illustration of allele frequencies for microsatellite marker D2S427 obtained by individual genotyping in the original sample set (left) and the replication sample set (right). Also shown are AST results with one-tailed p-values.



AST (allele 5): $\chi^2 = 5.40$
 $p = 0.010$



AST (allele 5): $\chi^2 = 3.51$
 $p = 0.030$

Average g individuals
 High g individuals

Figure No. 26: Illustration of allele frequencies for microsatellite marker D4S2460 obtained by individual genotyping in the original sample set (left) and the replication sample set (right). Also shown are AST results with one-tailed p-values.

4.4.5 Population stratification control

To check whether significant results generated during the study arise from population stratification, a method suggested by Pritchard and Rosenberg (1999) was applied ⁶. One modification to the method was that marker allele frequencies were estimated and did not come from individual genotyping of the sample.

A set of 37 randomly chosen microsatellite markers on chromosome 6 was used for this purpose. For each marker allele frequencies were estimated on pooled DNA samples from Stage 1, using the algorithm described in section 4.3.4. The most frequent allele for each of the 37 markers was then compared to all other alleles. This yielded a non-significant result ($\chi^2=22.914$, $p=0.966$), suggesting that population stratification does not contribute significantly to observed differences between high g and average g individuals (See Table 25).

Marker ID	Group	Allele 1 Counts	Allele 2 Counts	χ^2		Marker ID	Group	Allele 1 Counts	Allele 2 Counts	χ^2
D6S1697	Mid	90	112			D6S1644	Mid	83	119	
	High	97	105	0.488			High	78	124	0.258
D6S1598	Mid	82	120			D6S1717	Mid	75	127	
	High	91	111	0.819			High	82	120	0.510
D6S1955	Mid	70	132			D6S408	Mid	96	106	
	High	79	123	0.861			High	84	118	1.443
D6S1279	Mid	82	120			D6S1620	Mid	64	138	
	High	62	140	4.316			High	67	135	0.102
D6S1034	Mid	66	136			D6S413	Mid	130	72	
	High	67	135	0.011			High	130	72	0.000
D6S1721	Mid	82	120			D6S270	Mid	89	113	
	High	85	117	0.092			High	94	108	0.250
D6S1593	Mid	109	93			D6S1699	Mid	83	119	
	High	104	98	0.248			High	82	120	0.010
D6S285	Mid	83	119			D6S314	Mid	48	154	
	High	90	112	0.495			High	47	155	0.014
D6S1266	Mid	68	134			D6S1704	Mid	62	140	
	High	60	142	0.732			High	68	134	0.408
D6S1029	Mid	133	69			D6S1637	Mid	50	152	
	High	127	75	0.388			High	53	149	0.117
D6S1629	Mid	62	140			D6S1064	Mid	111	91	
	High	65	137	0.103			High	110	92	0.009
D6S1680	Mid	76	126			D6S425	Mid	73	129	
	High	67	135	0.877			High	60	142	1.894
D6S1641	Mid	119	83			D6S1612	Mid	82	120	
	High	134	68	2.379			High	91	111	0.819
D6S282	Mid	37	165			D6S411	Mid	113	89	
	High	31	171	0.636			High	108	94	0.250
D6S1650	Mid	64	138			D6S1719	Mid	68	134	
	High	65	137	0.011			High	64	138	0.180
D6S1628	Mid	63	139			D6S1027	Mid	97	105	
	High	57	145	0.427			High	93	109	0.159
D6S1625	Mid	90	112			D6S281	Mid	85	117	
	High	75	127	2.305			High	79	123	0.369
D6S1634	Mid	92	110			D6S1626	Mid	67	135	
	High	97	105	0.248			High	72	130	0.274
D6S1601	Mid	67	135							
	High	61	141	0.412						
						Total	df=37	ΣX^2	22.914	
								p	0.966	

Table No. 25: Population stratification analysis ⁶. Analysis was performed according to suggestion by Pritchard and Rosenberg (1999). Sum of individual χ^2 values (ΣX^2) and the overall p-value of randomly selected 37 chromosome 6 microsatellite markers are presented at the bottom of the table.

4.5 Discussion

A systematic genome wide screen with close to 2000 microsatellite markers was carried out in an attempt to locate some of the genetic loci contributing to the total variance in *g*, and it is probably the first association study to employ a DNA pooling technique on such a large scale.

The objective of the study was to test ~3500 microsatellite markers with average inter-marker distance of 1 cM. To be selected for the screen, markers had to be highly polymorphic and to have high heterozygosity in order to increase the power to detect linkage disequilibrium with a trait locus¹⁵⁷, and preferably be 1 cM away from the first neighbouring selected microsatellite. Less than 3000 markers fitted those requirements.

Initially, the main focus was on finding optimal PCR conditions at which most markers would successfully amplify at first attempt. Once a successful protocol was established, the focus shifted to a non-templated addition of a nucleotide at the 3' end, which lead to serious problems of misinterpreted AIPs, falsely inflated simulated p-values and non-reproducible results. This was mainly resolved by introducing a universal and amenable to high throughput screening post-PCR Klenow fragment treatment of all markers, regardless of their predisposition to such PCR artefacts. In addition to increasing the quality of obtained AIPs, standardised protocols insured less variation in results when tests were carried out by different researchers, and it greatly contributed to a considerable speed at which the screening process was performed.

Although such measures ensured speed and reproducibility, they resulted in a relatively large number of markers which failed to amplify (38%) since optimisation of individual markers was not carried out. In consequence, only a 2 cM average marker spacing was achieved instead of the aimed tighter 1 cM grid.

Unlike markers with uncertain PCR behaviour selected by our laboratory, commercially designed genome screening panels of microsatellites (Applied Biosystems),

with incorporated "pig-tails" at their 5' ends, proved to be much more successful, with only 13% failing to generate good quality AIPs. For this reason, research groups planning systematic genome screens may wish to consider readily available marker panels that have been designed and optimised to produce accurate and reliable genotyping results under standard conditions. Greater initial cost may soon prove to be a good investment since, by having fewer dropouts, better genome coverage is ensured and fewer marker replacements required.

Centralised DNA pooling had an important role in reducing between-research groups variation by providing DNA templates of the same composition to both centres involved in the study. The quality of DNA comprising the pool proved to be of great importance. Old and already fragmented DNA can lead to rapid deterioration of the pool composition and make such pools unrepresentative of the pooled sample. The IQ QTL association project was fortunate in that its almost unlimited DNA supplies came from permanent cell lines established for each individual participating in the first stage of the study. High quality DNA derived from the cell lines resulted in DNA pools still highly replicable and truly representative of the sample even five years after their construction. In addition, equal contribution of DNA from all individuals comprising the pool is of utmost importance for studies based on estimated allele frequencies using pooled DNA¹⁹⁵. In this project, Pico-Green DNA quantification method was used.

Out of 1847 markers for which AIP images were generated in the first stage, only 1 (D4S2460) survived the final stage. The London team had access to sample of 196 parent-offspring trios, for which children were selected for high g ²¹³. The surviving marker was genotyped individually in the trios sample for Transmission Disequilibrium Test (TDT) analysis¹¹¹. The target allele 5 for D4S2460 was less frequent in the high g group in both original and replication case/control sets, while the other alleles were expected to be preferentially transmitted in high g individuals. Not only that the TDT result was not significant, but it also was in the wrong direction. That is, the target

allele was more frequently transmitted to high g offspring than not (55% vs. 43% respectively).

One reason for the TDT's failure to replicate the positive finding from the case/control sample may be population stratification. However, all subjects used in the study were classified as Caucasian, which means that population stratification is an unlikely contributor to observed allele frequency differences between average g and high g individuals¹¹². In addition, the genomic control analysis using pooled DNA data showed no evidence of population stratification ($\chi^2=22.914$, $p=0.966$).

It may seem that 1847 markers is a large number, especially when compared to 300-400 markers usually considered as sufficient for a good genome coverage in linkage based studies. Nevertheless, this number is far from the ideal of 35,000 – 250,000 bi-allelic markers thought to be necessary for a genome screen for association studies based on linkage disequilibrium^{132, 133, 146}, not to mention a daunting 1 million figure if Kruglyak's predictions prove to be correct¹¹⁷.

The power to detect QTL contributing to high g drops steeply with distance from the nearest marker. When D' is 1.0, the power to detect a QTL that accounts for 5%, 2.5% and 1% of the total variance is, respectively, 100%, 93% and 56%. When D' drops to 0.50, the power is reduced to 73%, 42% and 19% respectively. The average inter-marker distance for markers used in this study is 1.5 Mb, which means that no QTL should be further than 750 kb from the nearest marker. However, the average distance over which an LD between a microsatellite and a trait locus can be detected in outbred population is suggested to be between 300 and 500 kb^{135,214,215}. That leaves a relatively large proportion of the genome unaccounted for in this study. It would not come as a great surprise then if true associations with high g were missed. Low marker density problem may not be solved simply by testing additional several thousand microsatellite markers, because the power and accuracy of LD mapping with microsatellites becomes small if genetic distance between a microsatellite and a trait locus is very short¹⁵⁶. A more reasonable following step would be to genotype densely

distributed SNPs, in regions around microsatellite markers that gave positive signals in the study and in candidate genes/gene systems and their regulatory regions. Assuming that a coding SNP is the QTL, focusing efforts on them may further increase chances for detection of loci contributing to g .

Another possible reason for so few markers surviving to the final stage could be the over-stringent criteria imposed by the multiple stage study design. This is particularly true for the Stage 1, where the power to detect a QTL of small effect size (accounting for 1% of the total variance), even in the presence of the complete LD, is only 56%.

One way to approach this problem could be to relax requirements for the passage through to the next stage by allowing all markers with $p < 0.1$ to qualify for the Stage 2 which should halve Type 2 error. Even though lower stringency at the initial stage would result in doubling the Type I error (0.00025), the study design would still provide adequate protection against false positive results in a scan of 2000 markers. Reduction in the number of false negative results can also be achieved by using a more powerful sample. Larger sample sizes would have provided more power, but size changes were not a feasible option since all DNA collection aimed for the Stage 1 ended well before the molecular genetic study started. Choosing individuals with phenotypes further away from the mean ("super high" g) was another possibility, which actually did take place, but only in time for the replication stages. It has to be said, though, that permanent cell lines were not created from "super high g " individuals and that the very limited amount of DNA available rendered the sample set unsuitable for the initial screen of ~2000 markers. In addition, a more powerful sample set for replication stages has its advantage since replication of original findings requires more power.

Selection of the most extreme individuals in a phenotypic distribution is known to increase the power of a study¹⁹⁵, but in some instances the reverse may happen if a phenotypically different population subset is used for replication stages. When a phenotype-like general cognitive ability is influenced by a large number of genes and environment, individuals may show extreme phenotypes due to different sets of QTLs

and/or due to environmental factors ²¹⁶. This could be one of the reasons why out of 108 microsatellite markers found associated with high *g* in the original sample set, only 4 generated significant p-values in the more powerful replication set.

General cognitive ability is a complex trait with undetermined number of loci contributing to phenotypic variation. There is a real possibility that this number is large and that every individual QTL accounts for less than 1% of the total variance. Sample sizes used in the study do not provide sufficient power to detect such loci. Significantly larger sample sizes (~1000 or more) would increase the power, a benefit which does not necessarily have to come at a greater cost if DNA pooling is employed.

5 INVESTIGATING CANDIDATE GENES FOR *g*

High cognitive ability has been associated with various phenomena ranging from decreased glucose intake caused by more efficient synapses⁵², to faster nerve conduction^{49, 48}, to quicker reaction times^{49, 50, 51}, to better working memory and more effective executive functioning¹⁰. The correlation between psychometric *g* and these physical variables means that in some way *g* is connected with underlying biological systems. We now face a complex task of discovering the nature of these diverse biological systems and the way each of them shapes our cognitive ability.

5.1 *Apolipoprotein E*

5.1.1 Introduction

There are two processes thought to be essential for cognitive performance: learning and memory³³. A poor cognitive performance, often associated with a loss of synapses, is readily observable as part of the normal ageing processes²¹⁷. While most healthy elderly people will show some degree of forgetfulness and will perform less well than younger persons, some of them will demonstrate a significant cognitive decline. By the age of 85, about one third of the population will be affected by major impairments in intellectual abilities, and Alzheimer's disease (AD) is the commonest cause²¹⁸.

The $\epsilon 4$ variant of the apolipoprotein E (apoE) gene is one of the widely accepted factors which increase the risk and decrease the age of onset of Alzheimer's disease in direct relation to the number of $\epsilon 4$ alleles expressed^{219, 218}. In addition, more than

20% of the overall risk of impaired cognitive function in the general population is attributable to the ϵ 4 variant ²²⁰. A three-year longitudinal study of elderly men (mean age=75.1, SD=4.6) conducted in the Netherlands has shown that ApoE4 is also responsible for a more rapid cognitive decline in men with two copies of the allele, while in those with only one copy of the allele, the decline, although slightly slower, is still faster than for non-ApoE4 carriers ²²⁰. Even though some studies failed to show such relationship ²²¹, many others confirmed it and strengthened the hypotheses that poor cognitive performance is associated with the apoE ϵ 4 allele ^{222, 223, 224}.

5.1.2 Molecular biology of apolipoprotein E – general features

Apolipoprotein E is a plasma protein involved in metabolism of cholesterol and triglycerides. It is present in very low-density lipoprotein (VLDL) and high-density lipoprotein (HDL), and serves as a ligand for uptake of these molecules in the liver and other organs. Apart from mediating extracellular cholesterol transport, apolipoprotein E is known to regulate multiple additional metabolic pathways, mainly through its interactions with different receptors and through binding other proteins ¹¹. Much research is now directed at identifying these metabolic pathways in the hope that some of these could explain the development of neurologic and vascular disorders.

The apolipoprotein E gene is positioned on chromosome 19q13.32. Genomic sequence containing four transcribed exons is 3590 bp long (Figure No. 1). The expression of the gene gives rise to a 34-kDa protein made up of 229 amino acids. The protein is present in three common isoforms: apoE2, apoE3 and apoE4 ²²⁵. The isoforms have identical amino acid sequence (Figure No. 27) except for amino acids 112 and 158, where Cys112 and Arg158 (apoE3) are replaced by arginine at 112 (apoE4), or cysteine at 158 (apoE2). The amino acid changes determine the binding properties of

the molecule to receptors and other proteins, and are ultimately responsible for isotype-specific roles in pathology of neurologic ^{219, 226, 227}, and common vascular diseases ^{228,}

229

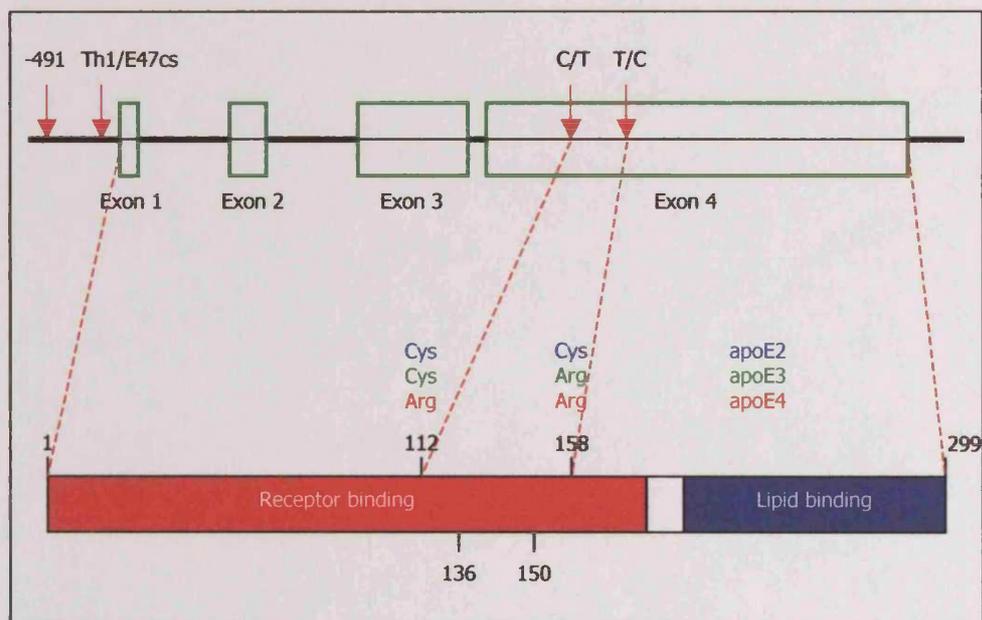


Figure No. 27: Molecular structure of the ApoE gene (top); three common isoforms of the apoE protein and its functional domains (bottom); binding of apoE to the LDL family of receptors requires conserved α 136 – α 150 region of the receptor binding domain ¹¹.

Functionally, there are two distinct domains of the protein: a receptor-binding and lipid-binding domain (Figure No. 27). The polymorphic sites determining apoE isoforms are located in the receptor-binding domain and, although not directly within the conserved region (α 136-150), they influence the binding affinity for LDL receptors. Apart from a well known role in mediating cellular uptake of lipoproteins, the LDL receptors have recently been shown to be involved in cellular signalling and neuronal migration during brain development ^{230, 231}. The variation in affinity with which the

apolipoprotein E isoforms bind to the LDL receptors is, therefore, likely to cause both disturbances in lipoprotein metabolism, leading to atherosclerosis¹¹ and to affect the efficiency of cellular signalling pathways²³⁰.

5.1.3 ApoE in neuronal maintenance and repair

Apolipoprotein E is the predominant type of apolipoprotein in the brain, and is involved in cholesterol metabolism in this organ. It is expressed by astrocytes, microglia and, to a lesser extent, neurons²³² in both normal and AD brain. Ignatius *et al.* (1986) were among the first to demonstrate that apoE levels are up-regulated in response to brain injury²³³. More recent studies further suggest that apolipoprotein E helps protect the brain against injury²³⁴, but the effect seems to be isoform specific. Possession of apoE4 variant increases the risk of an unfavourable outcome after a traumatic brain injury^{235, 236}, whereas apoE3 shows a neuroprotective effect²³⁷.

The exact mechanism underlying these processes is still unclear, but it could involve the influence of apoE on neuronal repair through its effects on neurite outgrowth and cytoskeletal stability²²⁵. It seems that apoE might also be necessary for maintenance of neuronal integrity during normal ageing of the CNS^{238, 239}.

In vitro studies have demonstrated that apoE4 and apoE3 have differential effects on neuronal sprouting. Addition of apoE4 to the culture medium inhibits neuronal outgrowth, while apoE3 variant enhances neurite extension^{240, 241}. These findings were later confirmed by a number of *in vivo* studies using transgenic mice expressing human apoE4 and apoE3 variants^{242, 239, 243}. The adverse effect of apoE4 on neurite outgrowth was attributed to its ability to destabilise the neuronal cytoskeleton, especially the micro-tubular system, which is known to be essential for neuronal development and remodelling²⁴⁴.

Buttini *et al.* (2000) showed that detrimental apoE4 effects are dose-dependent and that defects in neuronal sprouting are further exacerbated with increased levels of apoE4²⁴⁵. Not only is apoE4 defective in supporting neuronal extension, but it also acts as a dominant negative factor interfering with the beneficial effects of apoE3^{245, 232}.

In experiments using human ApoE transgenic mice, Cambon *et al.* (2000) have shown that apoE4 is associated with a significant decline in a synapse per neuron ratio²³⁹. The effect becomes more apparent later in life when neuronal repair is required. It can be argued that the ϵ 4 isoform is less efficient in reparation of damage incurred by natural ageing processes such as synaptic loss. The loss of synapses and/or replacement by aberrant ones are likely to disrupt hippocampal circuitry, which could affect the storage and retrieval of information in the hippocampus and, quite possibly, result in cognitive impairment²⁴⁶.

5.1.4 Transcriptional activity of the ApoE gene

Even though studies described so far show that the ApoE ϵ 4 allele increases the risk of developing Alzheimer's disease and that it has a deleterious effect on neuronal maintenance and repair, the studies do not explain why a relatively large proportion (~50%) of aged homozygotes for ϵ 4 allele¹⁷ do not develop cognitive impairments.

It is possible that polymorphisms within the transcriptional regulatory region of the ApoE gene influence these phenomena. To date, five polymorphisms have been described in the ApoE promoter region (see Figure No. 1 and 2)^{16, 17}. Two of those seem to play a role in AD. Lambert and colleagues characterised a G>T mutation at –186 bp, named Th1/E47cs, located in a consensus sequence of the basic helix-loop-helix transcription factor, E47¹⁷, while Bullido *et al.* described an A>T polymorphism at –491 bp²⁴⁷.

Both groups found the A allele of the -491 bp site to be an independent risk factor for the disease. *In vitro* studies have shown that the -491A allele is associated with higher levels of ApoE promoter activity than the T allele²⁴⁷. The authors hypothesised that the homozygous AA genotype, but not the heterozygous AT, was responsible for the increase in risk for Alzheimer's type dementia. This hypothesis was supported in North American and Caucasian French samples^{247, 248}.

```

-1000 gagccaggattcacgccctggcaatttgactccagaatcctaacctaac
-950 ccagaagcacggctcaagcccctgaaaccacaatacctgtggcagcca
-900 gggggagggtgctggaatctcatttcacatgtggggagggggctcccctgt
-850 gctcaaggtcacaaccaagaggaagctgtgattaaaacccaggtcccat
-800 ttgcaagcctcgacttttagcaggtgcatcactgttcccaccctcc
-750 catcccacttctgtccagccgcctagcccacttctttttttctttt
-700 ttgagacagtctccccttctgtgaggctggagtgagtgccgagatctcg
-650 gctcactgtaacctccgcctcccgggtcaagcgattctcctgcctcagc
-600 ctccaagtagctaggattacagcgcccgccaccacgcttgctaactt
-550 ttgtatTTTTtagtagagatggggttccaccatgttggccaggctggctc
-500 aaactcctg ccttaagtgattcgcccactgtggcctccaaag ggctgg
-450 gattacagggcgtgagctaccgcc cccagcccctcccattcccacttctgtc
-400 cagccccttagccctacttctttctgggatccaggagtccagatcccca
-350 gccccctctccagattacattcatccaggcacaggaaaggacagggtcag
-300 gaaaggaggactctgggcggcagcctccacattccccttccacgcttggc
-250 cccagaatggaggagggtgtctgtattact gggcgagggtgcctccctt
-200 cctggggactgtgg ggggtgtcaaagacctctatgccccacctcctc
-150 ctccccttgccctgctgtgcctggggcagggggagaacagcccacctcgt
-100 gactgggggctggcccagcccgcctatccctgggggagggggcgggaca
-50 gggggagccctataattggacaagtctgggatccttgagtcctactcagc

```

Figure No. 28:

1kb region upstream from the ApoE transcriptional start site; five characterised polymorphisms are presented in red colour; two of the polymorphisms found to be associated with AD, -491A/T and -186G/T (Th1/E47cs) are boxed.^{16,17}

The Th1/E47cs mutation has a more complex theory behind it. Epidemiological studies have shown that its adverse effects are most pronounced in $\epsilon 3/\epsilon 4$

heterozygotes, who are at the same time heterozygous for the Th1/E47cs polymorphism¹⁷. Lambert *et al.* pointed that individuals heterozygous for both loci will have one of two haplotypes with the $\epsilon 4$ allele combined with either Th1/E47cs T or G allele. In their sample of 152 $\epsilon 3/\epsilon 4$ cases and 91 $\epsilon 3/\epsilon 4$ control subjects, AD cases were estimated to have $\epsilon 4$ more frequently combined with the Th1/E47cs T allele than controls (70% vs. 58%) resulting in an estimated overall AD risk of 1.7¹⁷. This was in accordance with their previous data which suggest that the $\epsilon 3/\epsilon 4$ individuals will have increased risk of developing AD because they expressed $\epsilon 4$ allele at a greater level²⁴⁹. Thus, the Th1/E47cs mutation in the APOE promoter may lead to preferential expression of the $\epsilon 4$ allele in $\epsilon 3/\epsilon 4$ individuals²⁴⁸, resulting in overall favouring of potentially deleterious action of the E4 isoform.

5.1.5 Aims

The findings reviewed above suggest that genetic variation in APOE modifies risk for cognitive decline with age. However, another hypothesis is that variation in APOE might influence cognitive function rather than influencing decline. Therefore, we investigated the influence of apolipoprotein E variations on general cognitive ability, g , in a young sample. A possible role of the differential expression of apoE alleles in cognitive performance was examined by observing two polymorphisms in the ApoE promoter region, Th1/E47cs and -491AT, which were repeatedly found to be associated with AD.

5.1.6 Materials and Methods

5.1.6.1 Sample

See General Materials and Methods

5.1.6.2 Genotyping of ApoE, Th1/E47cs and -491 AT polymorphisms

The ApoE genotypes were produced by PCR amplification of 8ng/ μ l DNA samples. 5'tccaaggactgcaggcggcga3' was used as a sense primer, and 5'acagaattcgccccggcctgtactactgcca3' as an antisense primer in the following conditions: 3min at 94°C, (30s at 94°C, 30s at 66°C, 1min at 72°C) x 35, 10min at 72°C. Restriction fragment length polymorphism (RFLP) analysis was performed by restriction digestion of the PCR products with *Cfo*I (2U per reaction, at 37°C for 16 h)²⁵⁰.

Th1/E47cs polymorphism was typed by nested PCR followed by double restriction digestion with *Hpa*II and *Eco*M. The primers used for PCR reactions were as follows: forward - 5'caaggtcacacagctggcaac3'; reverse - 5'tccaatcgacggctagctacc3'. First, they were used to generate a 1423 bp fragment spanning -1017 to +406 relative to the transcriptional start site¹⁶. Then, the product was used as a template to amplify smaller 234 bp fragment containing the polymorphic site. Because the polymorphism itself did not cause a change in restriction sites, the forward primer was designed to create a restriction site (forward: 5'agaatggaggagggtgtccg3'; reverse: 5'actcaaggatcccagactg3'; the underlined letters show positions of mismatched bases).

The -491 AT polymorphism was genotyped using the same nested PCR protocol as for the Th1/E47cs. With the larger fragment serving as a template, the smaller fragment with a forced restriction site was amplified using the primers: forward, 5'tgttgccaggctgtttaa3'; reverse 5'cctcttctgaccctgtcc3' (the underlined letters show

positions of mismatched bases). RFLP was analysed by double restriction digestion with *Dra*I and *Alu*I restriction enzymes.

5.1.6.3 Statistical analysis

Genotype frequencies for ApoE were analysed by $2 \times 5 \chi^2$ using the CLUMP program²⁵¹. Genotype and allele frequencies for Th1/E47cs and -491 polymorphisms were analysed by $2 \times 3 \chi^2$ using MINITAB 10.5 (1995). The haplotype analysis was performed using the EH+ program²⁵².

5.1.7 Results

Restriction isotyping of apolipoprotein E was performed by polymerase chain reaction (PCR) amplification of a discrete genomic region followed by *Cfo*I restriction digestion. The size of the uncut amplified product was 227 bp. Polymorphic sites determining apoE isoforms were easily recognised by either the presence or the absence of cutting sites for *Cfo*I restriction enzyme (see Figure No. 29 for graphical presentation of ApoE *Cfo*I restriction map)²⁵⁰.

Genotypes were obtained for 97 high *g* individuals++ and 98 controls (see Picture No. 1). The genotype and allele frequencies are shown in Table No. 26. No association of any of the apoE isoforms with *g* was found ($\chi^2=2.877$, $p=0.237$), although the differences in allelic frequencies were in the expected direction for the high *g* vs. average *g* groups (82% vs. 77%, respectively for $\epsilon 3$, and 13% vs. 15% for $\epsilon 4$). Genotypes also showed no significant difference ($\chi^2=3.175$, $p=0.211$), although the

genotypic frequencies were also in the expected direction for the high and average groups (65% vs. 58% for $\epsilon_3\epsilon_3$ and 25% vs. 22% for $\epsilon_3\epsilon_4$).

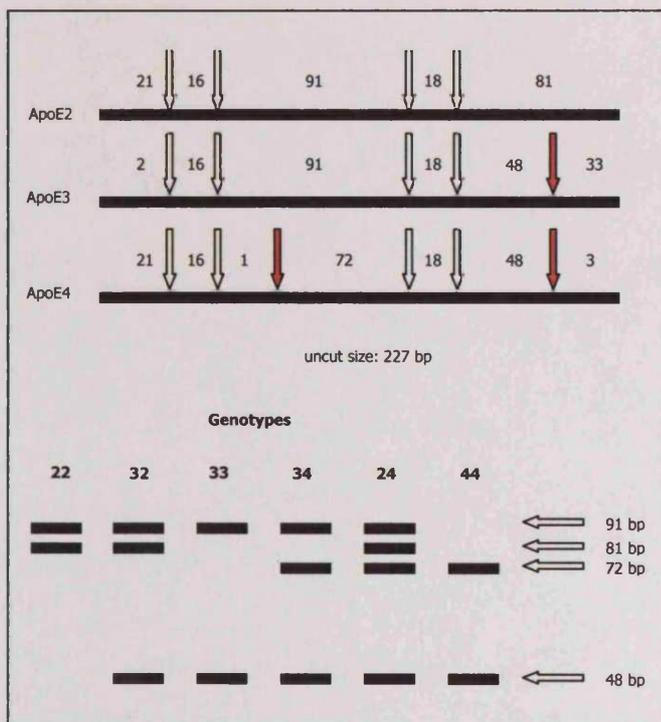


Figure No. 29:

ApoE *CfoI* restriction map; red arrows denote polymorphic restriction sites; the lower half of the figure shows restriction digests for different allele combinations and the way they would appear on agarose gel.

3 3 2 2 3 3 3 2 3 3 3 2 3 3 3
3 3 3 3 4 3 3 4 3 4 4 3 3 3 3



Figure No. 30:

ApoE *CfoI* restriction digestion products size separated on an agarose gel.

ApoE	Genotype n (f%) ^a						Allele n (f%) ^b		
	22	23	24	33	34	44	2	3	4
High <i>g</i>	0 (0.00)	9 (9.27)	0 (0.00)	63 (64.95)	24 (24.74)	1 (1.03)	9 (4.64)	159 (81.96)	26 (13.40)
Average <i>g</i>	0 (0.00)	14 (14.28)	3 (3.09)	57 (58.16)	22 (22.45)	2 (2.04)	17 (8.67)	150 (76.53)	29 (14.79)

Table No. 26: ApoE genotype and allele counts (n) and frequencies (f%)

a $\chi^2=3.17$; p=0.211

b $\chi^2=2.88$; p=0.237

To examine the possibility of the association between the apoE expression levels and cognitive ability, two polymorphisms in the transcriptional region of the ApoE gene (Th1/E47cs G/T and -491 A/T) were investigated.

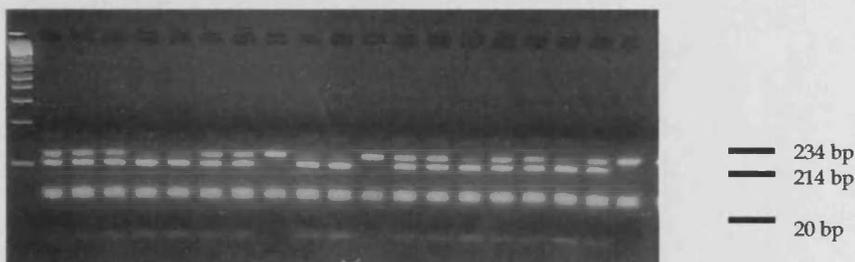


Figure No. 31: Th1/E47cs G/T polymorphism genotyping. Presented are *HpaII* and *EcoNI* restriction digestion products that are size separated on an agarose gel.

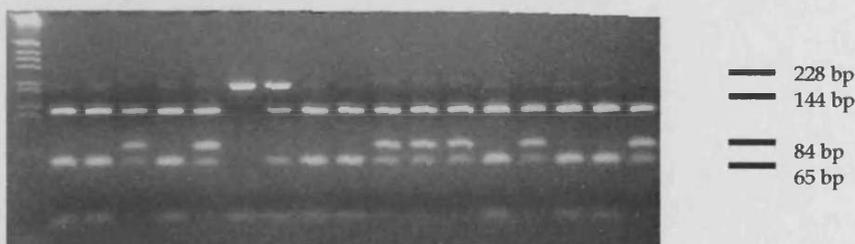


Figure No. 32: -491 A/T polymorphism genotyping. Presented are *DraI* and *AluI* restriction digestion products that are size separated on an agarose gel. An additional rare polymorphic site within the *AluI* recognition sequence was noticed. As a result, a variety of restriction products could be visualised on a gel. All possible combinations are described in a diagram below (Figure No. 33).

The allelic and genotypic frequencies of the two polymorphisms investigated (Table No. 27 and 28) were not different in the high *g* group when compared with the control group (Th1/E47cs genotypic frequency, $\chi^2=1.757$, $p=0.416$; Th1/E47cs allelic frequency, $\chi^2=0.389$, $p=0.533$; -491 AT genotypic frequency, $\chi^2=0.616$, $p=0.735$; ; -491 AT allelic frequency, $\chi^2=0.081$, $p=0.775$).

-491 A/T	Genotype n (f%) ^a			Allele n (f%) ^b	
	AA	AT	TT	A	T
High <i>g</i>	61 (64.89)	27 (28.72)	6 (6.38)	149 (79.26)	39 (20.74)
Average <i>g</i>	60 (61.22)	33 (33.67)	5 (5.10)	153 (78.06)	43 (21.94)

Table No. 27: -491 polymorphism genotype and allele counts (n) and frequencies (f%)

a $\chi^2=0.62$; $p=0.735$; $df=2$

b $\chi^2=0.08$; $p=0.775$; $df=1$

Th1/E47cs	Genotype n (f%) ^a			Allele n (f%) ^b	
	GG	GT	TT	G	T
High <i>g</i>	24 (25.53)	54 (57.45)	16 (17.02)	102 (54.26)	86 (45.74)
Average <i>g</i>	25 (26.32)	47 (49.47)	23 (24.21)	97 (51.05)	93 (48.95)

Table No. 28: Th1/E47cs polymorphism genotype and allele counts (n) and frequencies (f%)

a $\chi^2=1.76$; $p=0.415$; $df=2$

b $\chi^2=0.39$; $p=0.533$; $df=1$

The possibility that *g* might be influenced by differential expression of ApoE alleles was examined by analysis of the effect of promoter region variants in combination with ApoE isoforms. Differences in haplotype frequencies (Table No. 29 and 30) were not statistically significant for either of the promoter variants (Th1/E47cs – ApoE, $\chi^2=3.38$, $p=0.495$; -491 – ApoE, $\chi^2=5.18$, $p=0.300$; -491-Th1/E47cs-ApoE, $\chi^2=8.16$, $p=0.363$).

ApoE genotype	$\epsilon 2\epsilon 3$	$\epsilon 3\epsilon 3$	$\epsilon 2\epsilon 4$	$\epsilon 3\epsilon 4$	$\epsilon 4\epsilon 4$	n (f%)
High <i>g</i>						
AA	3	37	0	19	2	61 (64.89)
AT	2	21	0	4	0	27 (28.72)
TT	3	3	0	0	0	6 (6.38)
n (f%)	8 (8.51)	61 (64.89)	0 (0.00)	23 (24.47)	2 (2.13)	94
Average <i>g</i>						
AA	8	34	2	14	2	60 (61.22)
AT	5	20	1	7	0	33 (33.67)
TT	1	4	0	0	0	5 (5.10)
n (f%)	14 (14.29)	58 (59.18)	3 (3.06)	21 (21.43)	2 (2.04)	98

Table No. 29: ApoE and -491 A/T polymorphism combination

$\chi^2=5.18$; $p=0.300$; $df=7$

ApoE genotype	$\epsilon 2\epsilon 3$	$\epsilon 3\epsilon 3$	$\epsilon 2\epsilon 4$	$\epsilon 3\epsilon 4$	$\epsilon 4\epsilon 4$	n (f%)
High <i>g</i>						
GG	6	16	0	2	0	24 (25.53)
GT	3	34	0	17	0	54 (57.45)
TT	0	11	0	4	1	16 (17.02)
n (f%)	9 (9.57)	61 (64.89)	0 (0.00)	23 (24.47)	1 (1.06)	94
Average <i>g</i>						
GG	7	16	0	2	0	25 (26.32)
GT	6	27	3	11	0	47 (49.47)
TT	0	12	0	9	2	23 (24.21)
n (f%)	13 (13.68)	55 (57.89)	3 (3.16)	22 (23.16)	2 (2.11)	95

Table No. 30: ApoE and Th1/E47cs G/T polymorphism combination

$\chi^2=3.38$; $p=0.495$; $df=7$

5.1.8 Discussion

Alzheimer's disease is a complex disease likely to be caused by a number of genes working together with environmental factors²¹⁸. The $\epsilon 4$ allele of the apolipoprotein E gene is accepted to be one of the risk factors for the disease. Several recent studies have offered evidence that not only the presence of ApoE $\epsilon 4$ allele, but its differential expression are strongly associated with cognitive decline^{253, 248}. It can be argued that the $\epsilon 4$ isoform is less efficient in processes requiring reparation of damage incurred by brain injury and natural ageing processes such as synaptic loss. In these instances fewer new dendrites are formed and fewer lost synapses replaced, quite often with aberrant ones²⁴⁶. Therefore, it will not be surprising if these changes disrupted the finely tuned and complex signal transduction network, leading to less efficient cognitive functioning. This lead us to hypothesise that apolipoprotein E variations may also influence efficiency of cognitive functioning and general cognitive ability, g , in a younger sample. We investigated whether the different apoE isoforms and their differential expression plays a role in cognitive performance by observing polymorphisms in the ApoE coding sequence and its promoter region ($-491AT$ and $Th1/E47cs$).

No association between the ApoE isoforms ($\chi^2=2.88$; $p=0.237$), its promoter polymorphisms ($-491AT$: $\chi^2=0.08$; $p=0.775$ and $Th1/E47cs$: $\chi^2=0.39$; $p=0.533$) and general cognitive ability has been observed in the studied sample. The $\epsilon 4$ allele of the ApoE was not significantly less frequent in the high g group than in the control group and likewise the $\epsilon 3$ allele was not significantly more frequent in the high g group. The selected extremes design used in the present analysis has 100% power to detect QTLs that account for 5% of the variance and 93% power to detect QTLs that account for 2.5% of the variance ($p = 0.05$). Thus, the possibility that ApoE accounts for more than 2.5% of the variance could be excluded. However, it is possible that ApoE has a much weaker effect, as suggested by the allelic frequency differences for $\epsilon 3$ between

the two groups, although much larger samples would be needed to detect such a small effect.

Although there is evidence that the two ApoE promoter variants confer risk for developing AD in association with and independently of $\epsilon 4$ allele^{17,253}, the results that obtained in this study do not show that either the -491 A allele or the Th1/E47cs T allele is associated with g in the young healthy population.

In conclusion, apolipoprotein E is a major apolipoprotein constituent of the CNS where it plays an important role in injured and ageing brain by mediating neuronal remodelling. Its three isoforms have been shown not to be equally efficient in such processes. $\epsilon 4$, the least efficient of all is a well-established risk factor for AD which is the commonest cause of cognitive decline in elderly. However, this study suggests that apoE is not a factor with a measurable impact on general cognitive ability in the young healthy population.

5.2 Investigating Calcium/calmodulin dependent protein kinase II alpha (CaMKII- α) as a candidate gene for *g*

5.2.1 Introduction

5.2.1.1 Involvement of the hippocampus in learning and memory

Learning is manifested as a change in behaviour or in a probability of a particular response in the presence of a particular stimulus as the result of prior experience ¹⁰. Memory is the retention of the altered behaviour over a period of time.

Definitions of intelligence often include the ability to learn and to remember. The kind of learning and memory that is referred to as "spatial" is the ability to navigate effectively through an environment, integrating self-motion with prominent features of the external space ²⁵⁴. This ability has repeatedly been found to be one of the most heritable components of general cognitive ability, *g* (See Table No. 31) ^{7, 15, 19}.

Study	Heritability estimate	
	Spatial ability	Verbal ability
McGue & Bouchard (1989)	71%	57%
Pedersen (1992)	46%	58%
Ando (2001)	65%	65%

Table No. 31: Heritability estimates for specific cognitive abilities from three independent twin studies ^{7, 15, 19}.

The hippocampal formation is closely related to learning spatial relations. In fact, spatial memory is extremely hippocampus-dependent in rats ^{255, 256}. The hippocampus

has been implicated in the normal learning and memory in humans since the 1950s. Surgical removal of both medial temporal lobes in patient H. M.²⁵⁷ resulted in profound memory deficits, inability to acquire new information and to navigate in new environments²⁵⁸.

Impairments in spatial ability have also been recognised in patients with damage restricted to the right temporal lobe²⁵⁹ as well as in those with ischemic lesions limited to field CA1 of the hippocampus (See Figure No. 34)²⁶⁰.

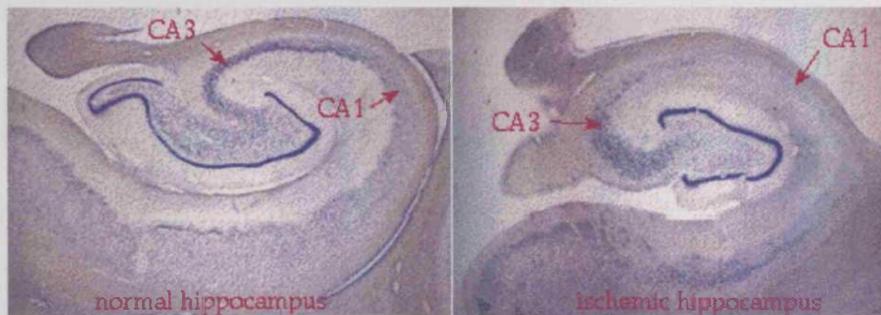


Figure No. 34:

A photomicrograph of a frontal section of the human hippocampal formation. The image on the left represents the normal hippocampus. The image on the right shows the ischemic hippocampus. Degeneration of CA1 and the absence of cell bodies (stained purple) is readily observable in the affected hippocampus¹².

Progressive learning and memory ability impairments and general cognitive decline are the main features of Alzheimer's disease, which is characterised by pronounced histopathological changes (extra-cellular plaques, intra-cellular neurofibrillary tangles and neuronal loss) in the CA1 region of the hippocampus (See Figure No. 35)^{261, 262, 263}.

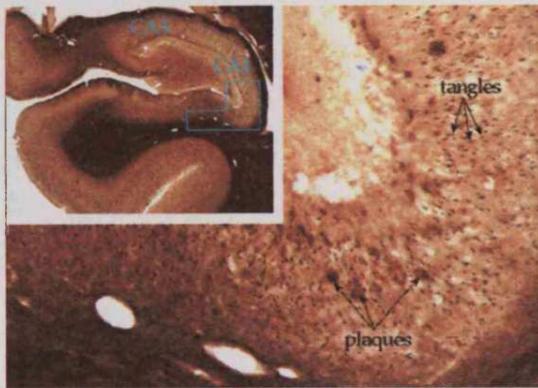


Figure No. 35:

A photomicrograph of a frontal section of the hippocampal formation in the brain affected by Alzheimer's disease. Pronounced histopathological changes in the CA1 region with visible neurofibrillary tangles and plaques (stained dark brown) ¹².

Taken together, these findings strongly indicate involvement of the hippocampus and its CA1 region in normal processes of human learning and memory.

5.2.2 Hippocampal anatomy and circuitry

The hippocampal formation is the folded archaecortex, which consists of two main features: the *dentate gyrus* and the *cornu ammonis* (CA), plus the subiculum (See Figure No. 36). The dentate gyrus and CA have the cortex organised into three layers. The subiculum is a transitional formation linking the hippocampus proper with the six layered entorhinal region of the neocortex ²⁶⁴.

The *perforant pathway* is a major afferent pathway bringing signals from the entorhinal neocortex to the hippocampus. Here it forms synapses with *granule cells* of the dentate gyrus and pyramidal cells of the CA3 region of the CA. CA3 cells also receive input from granule cells via *mossy fibers* ²⁶⁴.

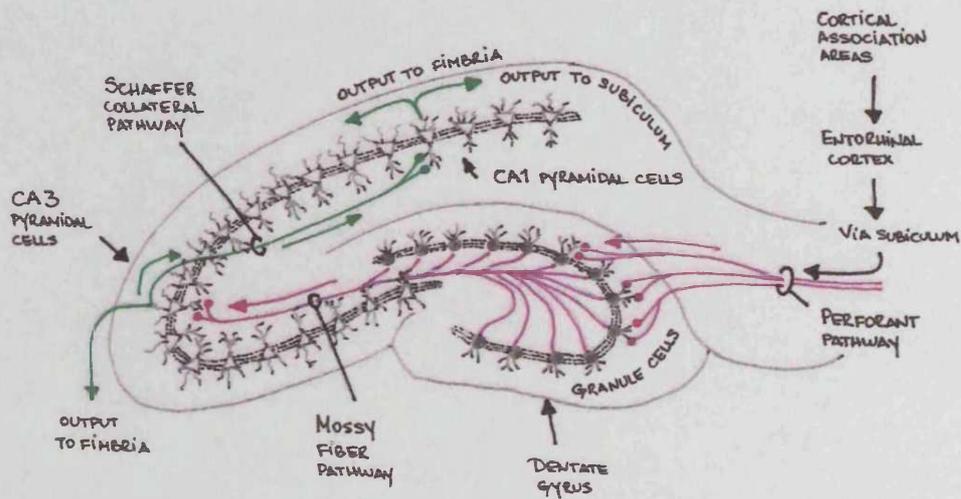


Figure No. 36: Diagram of the synaptic organization of the rat hippocampus (figure taken from <http://psych.fullerton.edu/swillis/amnesia.html>).

Axons of CA3 pyramidal cells form several efferent branches. Those that pass to the opposite side of the brain and make contact with the contralateral CA3 region are called *commissural fibers*. A proportion of the branches leaving the hippocampus will terminate in the hypothalamus and thalamus. Finally, some of the CA3 axons make synapses with neighbouring CA3 cells (*recurrent collaterals*) or with pyramidal cells in the CA1 region (*Schaffer collaterals*). CA1 cell axons make synapses in the subiculum and entorhinal cortex ²⁶⁴.

The three sets of excitatory synapses that are formed by the fibers of the perforant pathway terminating on granule cells and by the mossy fibers and Schaeffer collaterals making connections with the pyramidal cells are known as the *trisynaptic loop*, a formation that has been often referred to in studies on learning and memory mechanisms.

5.2.3 Neural basis of learning and memory

During development neuronal pathways are formed. Some of these connections remain unaltered and it is said of such pathways that they are "hard-wired". Pathways that are continuously rewired as a result of learning are referred to as "plastic" and the process itself as "plasticity"²⁶⁴.

Plasticity is observed across the animal kingdom, including animals with very rudimentary nervous systems. However, the less complex nervous system is less capable of making plastic changes in response to the environment because it has fewer neurons and fewer connections that are available for such modifications. In animals with more complex nervous systems, including humans, billions of neurons interconnect to form large networks *via* vast number of synapses, which continuously undergo changes in their strengths in response to various stimuli.

The idea of changes in synaptic strength as the basis of learning and memory is the core idea of the modern age neuroscience²⁶⁵. The first experimental demonstration of such changes took place in the early 1970s, when it was shown that repetitive activation of excitatory synapses in the hippocampus resulted in a long lasting increase in synaptic strength²⁶⁶. The observed phenomenon is termed "long-term potentiation" (LTP) and is usually defined as a persistent (hours) enhancement of an excitatory postsynaptic potential (EPSP) following brief high-frequency stimulation of afferent pathways²⁶⁶. A counterpart of LTP, long-term depression (LTD), is another event thought to be a component of a neural basis of learning and memory²⁶⁷. LTD can be induced by low frequency stimulation protocols, but its role in learning and memory is less well understood since only a small proportion of studies has focused on hippocampal LTD^{268, 269}.

All three types of excitatory synapses in the trisynaptic loop use glutamate as their neurotransmitter which can bind to metabotropic and ionotropic receptors. Ionotropic receptors can be further subdivided into two major types termed *α-amino-3-hydroxy-5-*

methyl-4-isoxazolepropionate (AMPA) and *N-methyl-D-aspartate* (NMDA) receptors. Depending on whether the activation of NMDA receptors is necessary for the inclusion of prolonged changes in synaptic strength, two distinct forms of hippocampal LTP can be recognised. While one of these, observed in the CA1 region, requires activation of NMDA receptors ²⁷⁰, the other, occurring at mossy fiber synapses in the CA3 region, is independent of NMDAR status and post-synaptic activity ^{271, 272}. Because the CA1 region of the hippocampus is regarded as a crucial neuro-anatomical formation for learning and memory, its NMDAR-dependent form of LTP has been more extensively studied in relation to these two behavioural processes.

The main feature of the CA1 type of LTP induction is its dependence on Ca^{2+} influx through activated postsynaptic NMDA receptors. The presynaptically released neurotransmitter glutamate binds to both types of postsynaptic ionotropic receptors, AMPAR and NMDAR. When the neuronal cell is close to its resting state, AMPA receptors are mainly responsible for influx of Na^+ and K^+ necessary for maintenance of the resting membrane potential. NMDARs are largely non-responsive at low frequency synaptic transmission, owing to voltage-dependent block by Mg^{2+} . Only strong afferent stimulation can produce sufficient postsynaptic depolarisation to remove the Mg^{2+} block and allow Ca^{2+} and Na^+ influx.

As described, for successful CA1 LTP induction two events have to occur simultaneously, namely presynaptic release of sufficient amounts of glutamate and sufficient depolarisation of the postsynaptic membrane to remove the Mg^{2+} block and open NMDAR channels. The mechanism involved in LTP induction was proposed more than 50 years ago by Donald Hebb, who postulated that a synaptic connection would be strengthened if the presynaptic and postsynaptic neurons were active simultaneously ²⁷³. Therefore, the coincidence detection property exhibited by NMDA receptors makes them crucial for memory formation under the basic principle of Hebb's rule.

5.2.4 Role of α -CaMKII in learning and memory

The sequence of events following LTP induction *via* NMDA receptor-gated Ca^{2+} influx is called LTP expression ²⁶⁴. One mechanism by which Ca^{2+} signals are converted into physiological changes in the neuron is through the activation of protein kinases. A family of serine/threonine protein kinases that is directly activated by Ca^{2+} is the Ca^{2+} /calmodulin-dependent protein kinase II (CaMKII) family.

5.2.4.1 Biochemical properties of CaMKII

CaMKII comprises a family of 28 similar isoforms derived from four genes (α , β , γ and δ) ²⁷⁴. CaMKII is a multi-unit holoenzyme (460 – 654 kDa) with a ring-like structure composed of 8 to 12 subunits ²⁷⁵. Most of the CaMKII activity in the forebrain comes from the hetero-multimers with an α : β isoform ratio of 3:1 ²⁷⁵. In addition, homomers composed of the α isoform have also been identified in the rat forebrain ²⁷⁶.

A typical CaMKII subunit possesses three domains: catalytic, autoregulatory and association (See figure No. 37). Of these, the catalytic domain is a substrate binding domain, but it is also capable of binding the autoinhibitory domain, which has amino-acid residues that mimic a protein substrate ²⁷⁷. The association domain is necessary for formation and assembly of a multimeric CaMKII holoenzyme.

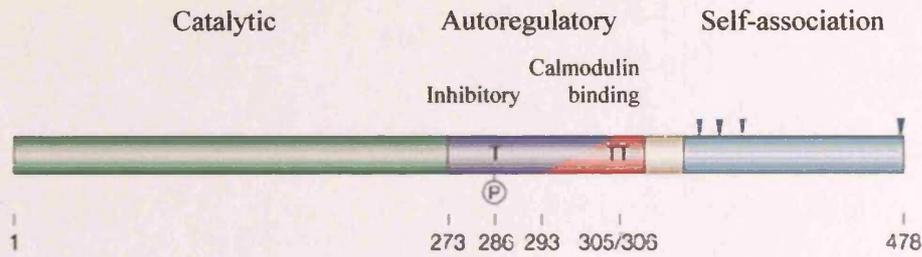


Figure No. 37: A typical CaMKII subunit possesses three domains: catalytic, autoregulatory and association domain. Denoted are phosphorylation sites (P) and amino-acid positions (1 - 478).

In the absence of Ca^{2+} /calmodulin ($\text{Ca}^{2+}/\text{CaM}$), the kinase is inactive owing to the autoinhibitory domain, which blocks the catalytic domain. $\text{Ca}^{2+}/\text{CaM}$ binding alters the conformation of the autoinhibitory domain and permits substrate access to the catalytic site. This activates a particular subunit, but will not lead to autophosphorylation of the enzyme unless another $\text{Ca}^{2+}/\text{CaM}$ is bound to a neighbouring subunit. This conformational change, triggered by $\text{Ca}^{2+}/\text{CaM}$ binding, exposes a site for phosphorylation at Thr²⁸⁶ in the adjacent subunit. Autophosphorylation of this residue further disrupts the autoinhibitory interaction with the catalytic domain and continues to produce such disinhibition in the absence of its activator $\text{Ca}^{2+}/\text{CaM}$.

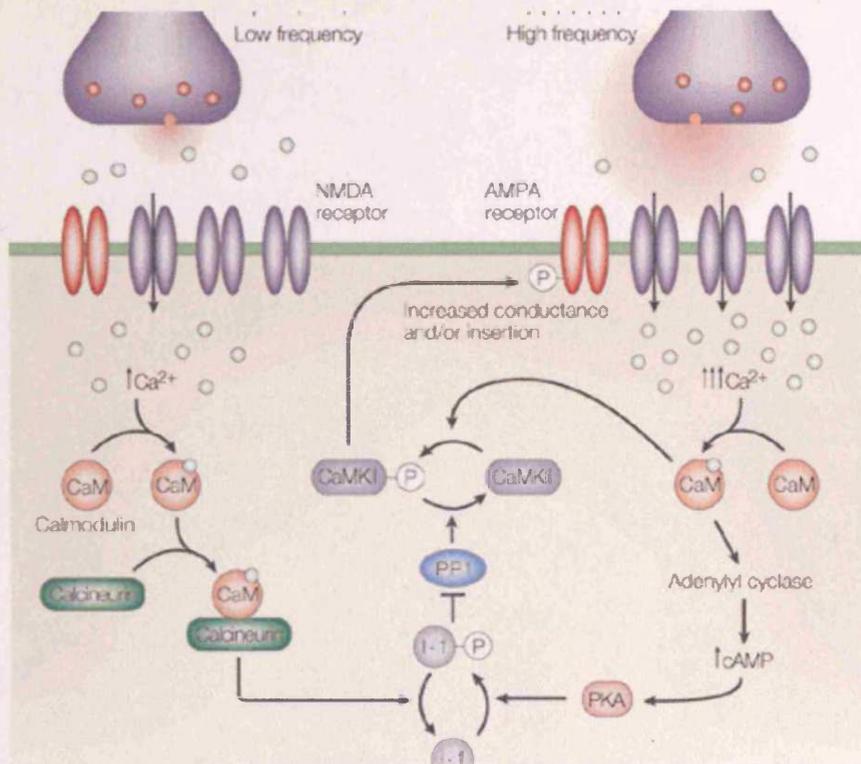


Figure No. 38: The Lisman model of kinase-phosphatase signalling in LTP⁵. Figure taken from¹³.

Autophosphorylation of Thr²⁸⁶ increases the affinity of CaMKII for CaM over 1000-fold, called "*CaM trapping*", by exposing residues Phe²⁹³ and Asn²⁹⁴, which are critical for a switch from low to high-affinity CaM binding²⁷⁸. This is followed by another wave of autophosphorylation, or "*burst phosphorylation*", of a number of residues within the Ca²⁺/CaM binding domain, which ultimately results in CaM dissociation and Ca²⁺/CaM independent enzyme activity. The inability of the phosphorylated and activated kinase to bind Ca²⁺/CaM is called "*CaM capping*"²⁷⁹. It has been suggested that the potential function of the burst phosphorylation might be to mark autonomously active subunits of CaMKII, so that these no longer compete for CaM binding and activation²⁷⁷.

The activity of phosphorylated CaMKII is also regulated through the action of protein phosphatase 1 (PP1), which is very effective in dephosphorylating Thr²⁸⁶ residue²⁸⁰. The activity of PP1 is suppressed during LTP in order to prevent rapid dephosphorylation and inactivation of CaMKII, whereby the suppression results from interaction of PP1 with inhibitor proteins, one of which is known as inhibitor 1 (I1)²⁸¹. When I1 is phosphorylated by protein kinase A (PKA), it inhibits PP1 and thus prolongs autonomous CaMKII activity²⁸².

5.2.4.2 CaMKII is sensitive to frequency of Ca²⁺ oscillations

In the brain, synaptic activity is marked by oscillations in intracellular Ca²⁺. Information on the stimulus is encoded by size, frequency and duration of these oscillations²⁸³. CaMKII is believed to be a central molecular mechanism involved in the decoding of such information.

Scientific evidence for such a proposal came from an *in vitro* study which showed that generation of autonomous CaMKII activity is sensitive to the frequency and size of Ca²⁺ spikes²⁸³. At lower Ca²⁺-spiking frequencies, kinase autonomous activity was low even though the autophosphorylation was evident. Stimulation with high frequency pulses (4.0 Hz) and/or with those of longer duration (500 to 1000 ms) resulted in a marked increase in enzyme autonomy. The observed CaMKII frequency-dependent autonomy occurs possibly because spike intervals are shorter at higher frequencies than the time required for complete dissociation of CaM²⁸³, which increases the probability of CaM molecules binding to two neighbouring subunits and subsequently reaches the "CaM trapped" state and full autonomous kinase activity.

The initial state of CaMKII activity also affects the way the enzyme responds to subsequent Ca²⁺ fluctuations. If several subunits are already autophosphorylated, the

enzyme sensitivity will shift to lower stimulation frequencies. In this way, CaMKII sets the stimulus frequency threshold at which the enzyme responds and determines the frequency threshold required to induce LTP or LTD ²⁸⁴. It was postulated that this enzyme could function as a molecular switch of learning and memory, whereby past Ca²⁺ influx is "memorised" in a form of various levels of autonomous enzyme activity and by regulated responsiveness to subsequent stimuli ²⁸⁴.

5.2.4.3 CaMKII is crucial for hippocampal LTP induction

The ability of the kinase to convert a transient Ca²⁺ signal into a long-lasting increase in enzyme activity has led to the suggestion that the kinase is directly involved in induction and expression of long-term potentiation ²⁸⁵. Evidence in support of the suggestion is overwhelming.

Research groups using various synthetic peptide analogues corresponding to the autoregulatory domain of CaMKII showed that extracellular application of such inhibitors can block LTP ^{286, 287}. However, the activity of these inhibitors was not specific ²⁸⁸ and, although a role of CaMKII in LTP and learning and memory could not be ruled out, it needed further scientific support.

In 1992, Silva and colleagues generated genetically modified mice that do not express the α -CaMKII ²⁸⁹, the isoform which is highly enriched in postsynaptic formations of hippocampus. These mice had no obvious neuroanatomical deficits, and despite intact postsynaptic NMDA receptor function, they had almost complete loss of normal LTP and displayed a severe deficit in spatial memory tasks ²⁸⁹.

By using regulated and targeted expression of the α -CaMKII gene, Mayford and colleagues (1996) addressed the question of whether the changes observed in genetically modified animals were due to an indirect effect on neuronal circuit

development. Using a forebrain specific promoter, they were able not only to time-control the gene expression, but also to restrict its expression to specific regions of the brain, particularly the hippocampus. Results of the study led to conclusion that the CaMKII signalling pathway is critical for memory processes in a manner that is independent of its potential role in brain development ²⁹⁰.

The importance of autophosphorylation of Thr²⁸⁶ of the α CaMKII in LTP and learning was demonstrated by introducing a point mutation into the α CaMKII gene, which blocked the autophosphorylation of the Thr²⁸⁶ residue without affecting its CaM-dependent activity. The mutant mice had no NMDAR-dependent LTP in the hippocampal CA1 area and showed massive reduction of spatial learning ability ²⁹¹.

5.2.4.4 CaMKII and LTP expression

The postsynaptic density (PSD) is a tiny, amorphous structure located beneath the postsynaptic membrane of synapses in the CNS ²⁹². It is usually seen as a thickening in the membrane caused by the accumulation of proteins, receptors and enzymes involved in the postsynaptic response to transmitter ²⁶⁴ and is believed to be the site of LTP expression ²⁶⁵.

It has been known for some time that CaMKII is localised in post-synaptic densities, where together with NMDARs it makes up a large proportion of the total PSD protein ²⁹³. Leonard *et al.* (1999) have shown that CaMKII recruitment to PSD is increased after the NMDAR-gated Ca²⁺ influx ²⁹⁴. Here the kinase becomes associated with NR1 ²⁹⁴ and NR2A/B subunits of NMDA receptors ²⁹⁵, which places CaMKII at an ideal position for activation by Ca²⁺ influx through NMDARs.

The anchoring of the CaMKII *via* NMDARs brings the kinase in close proximity to AMPA receptors, whose phosphorylation plays a central role in NMDAR-dependent LTP

expression²⁹⁶. The induction of LTP specifically increases the phosphorylation of the GluR1 subunit of the AMPA receptor²⁹⁶, the subunit demonstrated to be crucial for generation of LTP in hippocampal CA1 synapses²⁹⁷. This phosphorylation event mediated by CaMKII enhances AMPA receptor-mediated currents due to an increase in single-channel conductance which causes prolonged changes in synaptic strength in response to synaptic activity²⁹⁷.

In addition to modifying the conductance of existing AMPA receptors, CaMKII induces the insertion of new GluR1 containing AMPARs into the synapse, which causes further increase in AMPAR-mediated transmission²⁹⁸. The insertion of AMPA receptors leads to increase in PSD size and to eventual formation of multi-spine synapses and synapse multiplication²⁹⁹. Such structural changes in synapses triggered by CaMKII activity may well be a mechanism by which experiences are translated into memories.

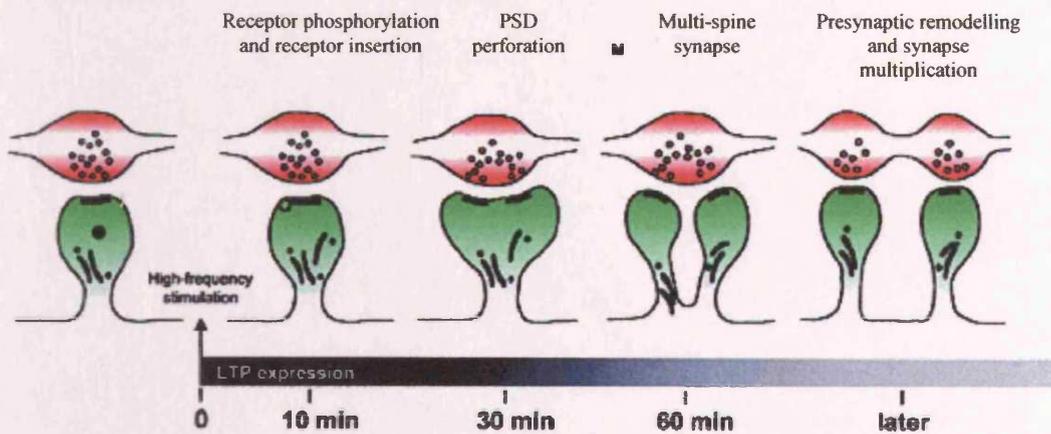


Figure No. 39: LTP expression mechanism (picture adopted from²⁹⁹): Activation of CaMKII results in phosphorylation of AMPA receptors and an increase in their single-channel conductance. In addition, AMPA receptors are inserted into the postsynaptic membrane, which leads to an increase in the size of the PSD. Through an unknown mechanism, some synapses use the expanded membrane area to generate multi-spine synapses. Retrograde communication triggers presynaptic structural changes, eventually increasing the total number of synapses²⁹⁹.

5.2.5 Organisation and expression of the α -CaMKII gene

The gene coding for α -CaMKII maps to human chromosome 5q33.1³⁰⁰. The gene is organised into 18 exons and 17 introns with an overall genomic size of 66641 bp. The gene is transcribed into a 4836 bp mRNA, which is alternatively spliced to produce two forms of the α -CaMKII. The longer transcript, variant A, is expressed in the brain where it is involved in LTP induction and expression, while the shorter transcript, variant B, lacks the characteristic kinase activity and is preferentially expressed in skeletal muscle³⁰¹.

The α -CaMKII mRNA is one of a limited number of mRNAs, including glutamate receptor family transcripts, shown to be present within the dendrites of neurons³⁰². Protein synthetic machinery has also been identified in dendrites, and it was clearly demonstrated that dendritic translation occurs independently of neuronal cell bodies³⁰³,³⁰⁴. Local protein translation in dendrites could be a means for delivering synaptic proteins to their sites of action in a fashion that could contribute to plasticity.

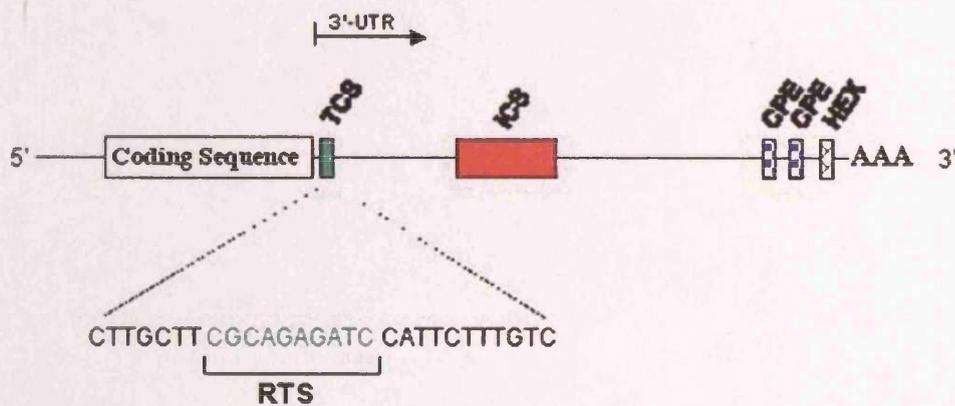


Figure No. 40: Functional map of the CaMKII 3'UTR, showing the cis-element regulation sites in the 3'UTR thought to be involved in dendritic targeting and translational control. Within the targeting cis-sequence (TCS) is an oligonucleotide sequence homologous to the RNA transport sequence (RTS) necessary for mRNA transport in oligodendrocytes. ICS denotes inhibitory cis-sequence that can prevent dendritic mRNA transport¹⁴.

While the forebrain specific expression of α -CaMKII is driven by the gene promoter region, the targeting of α -CaMKII mRNA to dendritic processes requires the presence of cis-acting elements in the 3' untranslated region (3'UTR) of the mRNA^{305, 306, 307}. Two particularly interesting cis-acting elements present in the 3'UTR of the α -CaMKII mRNA were described for the first time by Mori and colleagues (2000). α -CaMKII/Neurogranin dendritic localisation element (CNDLE) is a 27bp long domain located at the most 5' end of the 3'UTR, whose presence seems to be sufficient for dendritic targeting (See Figure No. 40). CNDLE was also identified in the 3'UTR of neurogranin mRNA, which is another transcript known to be targeted to dendrites³⁰⁷. The second regulatory element described is an inhibitory element that prevents mRNA translocation to dendrites. The authors have shown that synaptic activity marked by prolonged depolarisation can override this inhibition³⁰⁷. In addition to these two novel elements, two well defined sequences, the cytoplasmic polyadenylation element (CPE) and the poly-(A) addition site (HEX) are likely to be involved in activity-dependent translation of α -CaMKII³⁰⁸.

The importance of dendritic localisation of α -CaMKII mRNA and its local translation for hippocampal LTP was established by Miller and colleagues (2002)³⁰⁹. The group generated a mutant mouse whose protein-coding region of α -CaMKII was intact, but in which the dendritic localization signal in the mRNA (CNDLE) was disrupted and mRNA restricted to the soma. The removal of α -CaMKII mRNA from dendrites resulted in a remarkable reduction of the kinase in postsynaptic densities (PSDs), a reduction in LTP, and impairment in spatial memory³⁰⁹. These results demonstrate that local translation is important for synaptic delivery of the kinase and that local translation contributes to synaptic plasticity.

5.2.6 Aim of the study

Neuronal plasticity, as detected through a phenomenon of LTP, is a basis of learning and memory. CA1 region of the hippocampus is regarded as a crucial neuro-anatomical formation for spatial learning and memory. For successful CA1 LTP induction, activation of NMDAR *via* Ca^{2+} influx is necessary. One mechanism by which Ca^{2+} signals are converted into physiological changes in the neuron is through the activation of protein kinases such as α -CaMKII. The kinase is hypothesised to be directly involved in induction and expression of LTP through conversion of a transient Ca^{2+} signal into a long-lasting increase in enzyme activity and through mediating structural changes in synapses *via* insertion of new AMPARs. Dendritic localisation of α -CaMKII mRNA, mediated through *cis*-acting regulatory elements, and its local translation are crucial for hippocampal LTP and memory formation.

Therefore, we hypothesised that α -CaMKII isoform may influence *g* by affecting two of its major components, spatial learning and memory. We screened the gene for polymorphisms, including promoter, all exons, intron/exon boundaries, 3' and 5' UTRs. Detected polymorphisms were tested for association with general cognitive ability on high *g* and average *g* DNA pools using a primer extension method.

5.2.7 Materials and methods

5.2.7.1 Subjects

See General materials and methods section.

5.2.7.2 PCR amplification of the CaMKII- α genomic sequence

A complete genomic sequence for the CaMKII- α gene (Reference sequence ID: NM_015981), together with defined intron/exon boundaries, was obtained on the UCSC Genome Browser web site using a November 2002 sequence assembly version (NCBI Build 31) (<http://genome.ucsc.edu/>).

PCR primers (Table No. 32) were designed using the publicly available program Primer3³¹⁰.

PCR conditions were determined using the protocol described in General Materials and Methods section.

5.2.7.3 Denaturing High Performance Liquid Chromatography

Detailed description of the method is provided in the General Material and Methods chapter. PCR primers used for amplification of DNA fragments to be analysed by the method are listed in Table No. 32.

No.	Fragment ID	Forward primer sequence	Reverse primer sequence
1	CaMKII-5' flank	acggggcaatacctgtctct	agccttcgtcagcatcca
2	CaMKII-exo1	cgtttgagggtggtgc	tacaaggaatgctccccaaag
3	CaMKII-exo2	gtccctgagttgggtcacag	cccctgccataagtatcacc
4	CaMKII-exo3	aggcctcttctctttccac	ctccatccccagagaagaag
5	CaMKII-exo4	agcccagccctcttcagta	aagaggggaattcagccaag
6	CaMKII-exo5	cccaccttagtttctggag	gctggatacagtgactagggaga
7	CaMKII-exo6	caggtgagtgtaggtgctg	cctccagcctaacacagaga
8	CaMKII-exo7	gactgtcttggcgcttc	acagcatctccagccatacc
9	CaMKII-exo8	gctgttttctcagctccag	tacaggatgaccctggaaa
10	CaMKII-exo9	gtacgggaagcctgtggac	cccctgtgccagaactagag
11	CaMKII-exo10	gagtggatggcagtaggac	ttaccctgagaaccacagc
12	CaMKII-exo11	gatgtcaaccacacctcattaag	gtctgtccttccactagagcaa
13	CaMKII-exo12	gcttggtcttctctgaatg	ccatctagaactcaatcccctg
14	CaMKII-exo13	gtggatgctgctcgtcct	aaagcctaggtcgtgggtct
15	CaMKII-exo14	cagggtttcaggacagcat	aatgatgctcccagtggtg
16	CaMKII-exo15	tctgcatgaggtctcaggtc	taccocctgaacaacaatcc
17	CaMKII-exo16	tccaaggtccaatgagaag	gtaccaaccccagtccacac
18	CaMKII-exo17	gctgggagtgagaaagtga	ggcaacactcccaccttta
19	CaMKII-3'UTR-1	gactccatctaaagcggcag	gagagggacggacggatg
20	CaMKII-3'UTR-2	ggctgtaggctagaatgccc	cagaaatttggtataaaaaggc
21	CaMKII-3'UTR 3	tcttctctgtgctgacct	tccacagtccaaaaggaac
22	CaMKII-3'UTR 4	gtcacctgtggttattgggg	cctagtggatgtgccaaggt
23	CaMKII-3'UTR 5	tctgatgaccacagaaagc	cagttctgctccaaccaaca
24	CaMKII-3'UTR 6	gaagagccacgtcccag	tagccagacggttgctcctt
25	CaMKII-3'UTR-7	agggttcttggttccatcc	ggcactggagagtgacaca
26	CaMKII-3'UTR-8	aagagcagtgattgggggtg	ttgtctggtgagggtaggc
27	CaMKII-3'UTR 9	tgtcactctcaagtgccag	cttgaagggagacaggagg
28	CaMKII-3'UTR 10	taataaaaatccaaaccaagtca	tctggaactggaaactaacaanaac

Table No. 32: Oligonucleotide sequences used for PCR amplification of CaMKII- α fragments.

5.2.7.4 Fluorescent automated DNA sequencing using ABI PRISM® BigDye™

Terminators v 3.0

Detailed description of the method is provided in the General Material and Methods chapter.

5.2.7.5 Genotyping DNA pools using The SNaPshot™ Multiplex System

Detailed description of the method is provided in the General Material and Methods chapter.

5.2.7.6 Statistical analysis

Allele frequencies for identified SNPs were estimated from pooled DNA genotyping applying a correction formula, $A = A / (A + kB)$, described by Hoogendoorn *et al.*, where A and B are the peak heights of the primer extension products (representing alleles A and B in pools) and k is the mean of replicates of A/B ratios observed in a heterozygote²⁰.

Polymorphisms for which estimated allele frequencies were calculated were tested for allelic association with a general cognitive ability by a 2x2 χ^2 using MINITAB 10.5 (1995).

5.2.8 Results

5.2.8.1 Locating CaMKII- α promoter region

Transcription initiation in eukaryotes requires the presence of various combinations of short sequence elements positioned in the immediate vicinity of a gene, *cis*-acting, and at relatively constant distances from the transcription start site (TSS). These act as recognition signals for transcription factors and collectively constitute a promoter. A minimal DNA element necessary to produce an accurate initiation of transcription lies typically from -60 to +40 bp relative to TSS and is called a core promoter. Different types of genes will have different promoter compositions. Promoters of genes that are expressed at either a specific developmental stage or in a specific tissue and cell type will always include a TATA box. The TATA box (consensus sequence: TATA[A/T]A[A/T])³¹¹ is usually positioned ~25-30 base pairs upstream from the TSS. Many other genes, like housekeeping genes, are TATA-less. That is, they have CAAT box (CCAAT) at around -80 bp or a GC box (consensus sequence: GGGCGG) instead⁷⁰.

Even though the CaMKII- α promoter activity and its importance in localised forebrain expression was extensively studied in rats, no such reports on the human gene variant or any information on its core promoter sequence and exact location have been published so far.

Comprehensive screening of a gene for sequence variants would usually include detailed analysis of its promoter region. In attempt to localise one in the human CaMKII- α a number of publicly available databases and prediction programs were exploited. Results from the following three programs are presented:

1. PROSCAN - developed by BioInformatics & Molecular Analysis Section (BIMAS). The software predicts promoter regions based on scoring homologies with putative eukaryotic *Pol II* promoter sequences. No promoter regions were predicted in the first 500bp upstream from the TSS (see Figure No. 41).

```

Address  http://bimas.dcr.t.nih.gov/cgi-bin/molbio/proscan

Processed sequence:

1      CCCACTTAATCCCATCCCGTCTGCTACAAGGGCCCCACAGTTGGAGGTGG
51     GGGAGGTGGGAAGAGAAAAGATCACTTGTGGACAAAAGTTTGCTCTATTCC
101    ACCTCCTCCAGGCCCTCCTTGGGTCCATCAGGGGCTGCTGGGTCCA
151    TCCCACCCCCAGGCCACACAGGCTTGCAGTATTGTGTGCGGTATGGTCA
201    GGGCGTCCGAGAGCAGGTTTTCGCAGTGGAAAGGCAGGCAGGTGTTGGGGAG
251    GCAGTTACCGGGCAACGGGAACAGGGCGTTTGGAGGTGTTGCCATGG
301    GGACCTGGATGCTGACGAAGGCTCGCGAGGCTGTGAGCAGCCACAGTGCC
351    CTGCTCAGAAGCCCCGGGCTCGTCAGTCAAACCGGTTCTCTGTTGCACT
401    CGGCAGCACGGGCAGGCAAGTGGTCCCTAGGTTCCGGGAGCAGAGCAGCAG
451    CGCCTCAGTCTGGTCCCCAGTCCCAAGCCTCACCTGCCTGCCAGCGC
501    CAGGATGGCCACCATCACCTGCACCCGGCTTACGGAAAGGTACCAGCTCT
551    TCGAGGAATTGGGCAA

Proscan: Version 1.7
Processed Sequence: 566 Base Pairs
No promoter regions where predicted.

```

Figure No. 41: PROSCAN output file for the first 500 bp upstream from the CaMKII- α transcription start site.

2. **CISTER** - developed by ZLAB (Boston University, USA). Cister predicts regulatory regions in DNA sequences by searching for clusters of cis-elements³¹². No promoter or any other transcriptional regulatory elements were predicted in the first 500bp upstream from the TSS of the human CaMKII- α gene.

Cister

Results for CaMKII exon1 and 500 bp upstream (1 to 566 out of 566 nucleotides)

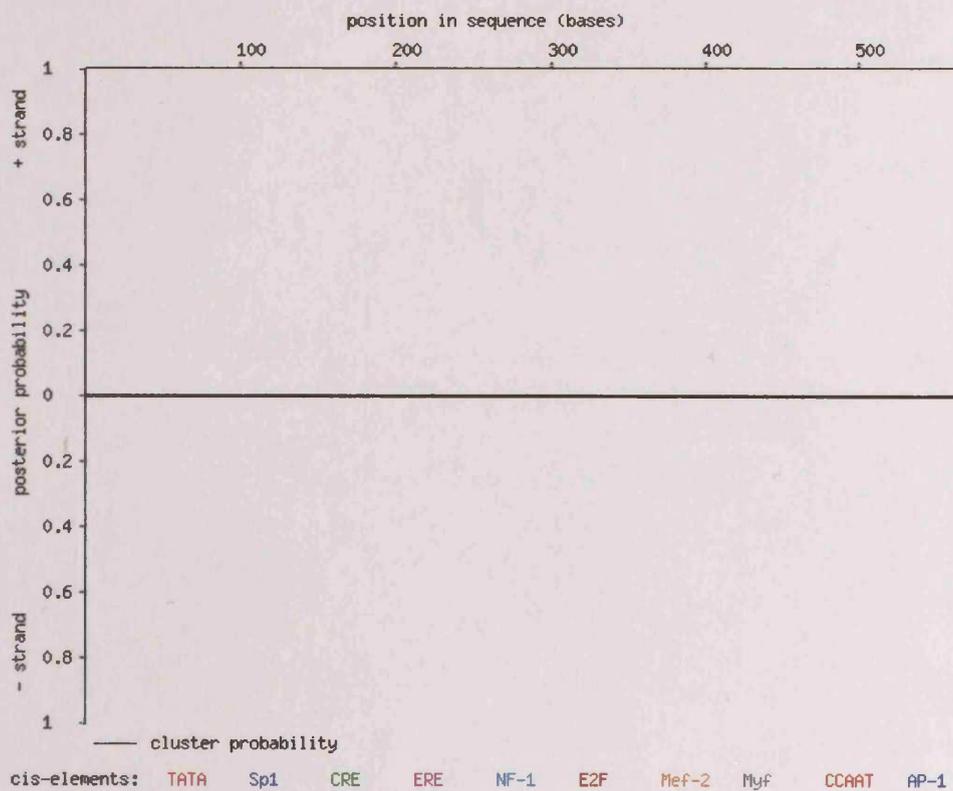


Figure No. 4: CISTER output file for Exon 1 and the first 500 bp upstream from the CaMKII- α transcription start site.

3. **NIX** - distributed by UK HGMP Resource Centre. NIX is a WWW tool to simultaneously view the results of running many DNA analysis programs on a particular DNA sequence. The analysis programs run include: GRAIL, Fex, Hexon, GENESCAN, Genemark, Genefinder, FGene, BLAST, RepeatMasker, tRNAscan *etc.* No promoter elements were predicted to exist in the region immediately upstream from the Exon 1 of the CaMKII- α gene. The only promoter element predicted was located downstream from the Exon 1 and was of a marginal prediction quality (displayed as a dark green triangle).



Nix Results for: cutpaste-8956

[Information on the results viewer](#)

Description: CaMK2A 5' flanking region and Exon 1

Sequence range: 1 to 1350 (Full sequence)

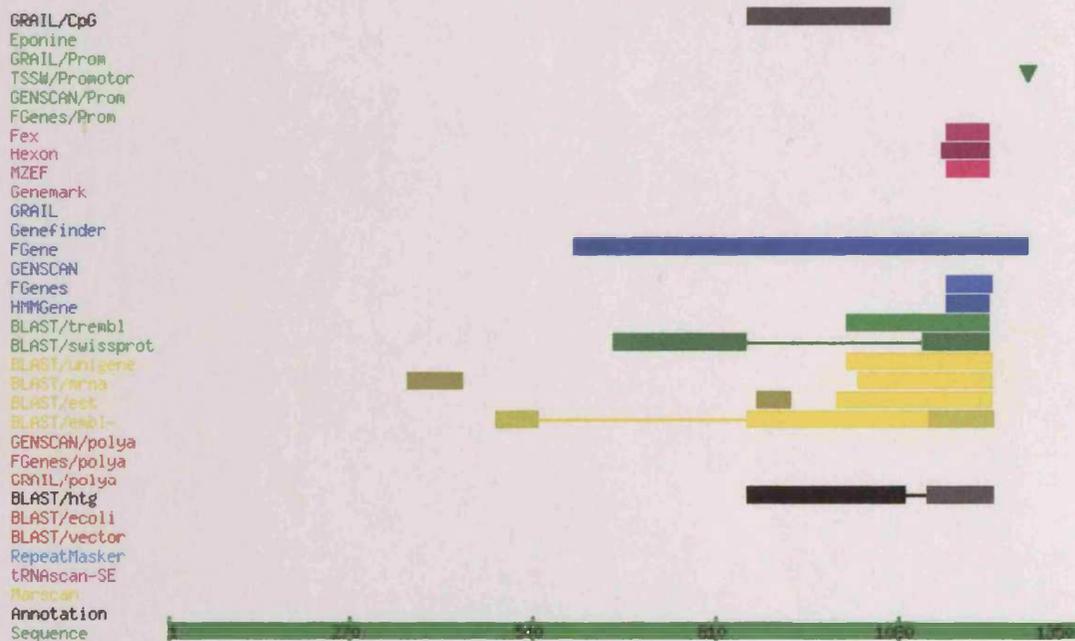


Figure No. 43: NIX output file for 1350bp including Exon 1 and the 5' flanking region of the CaMKII- α .

5.2.8.2 PCR, DHPLC and sequencing

A successful PCR amplification was achieved for most exons, putative promoter region and the whole length of the 3'UTR. Exon 12 is the only fragment for which the amplification assay was not successfully optimised even though different PCR conditions and two sets of primers were attempted, so that it had to be excluded from further assays and analyses.

For the purpose of mutation detection using DHPLC, each fragment was amplified on fifteen individuals from the high *g* group (95% power to detect polymorphisms with a minor allele frequency of 0.1). PCR products were denatured using the melt program (See General Material and Methods) and were subsequently run on a pre-programmed DHPLC machine. Amplified sequences containing exons 6, 7, 9, 15, 16, 17 and five amplicons from the 3'UTR exhibited changes in elution profiles. Profiles for all individual samples that subsequently underwent sequencing are presented in the Appendix section.

The nature of the polymorphisms and their respective positions were determined by sequencing (See Figure No. 44). In total, fourteen sequence changes were detected, three of which were insertion/deletions. Only one change was exonic (C/T change in exon 17, not causing an amino acid change), 5 were located in the 3'UTR while the remaining 8 were intronic.

Figure No. 44: Polymorphisms detected by dHPLC method and characterised by sequencing for 11 PCR fragments.

GATGCCAGgtgagtgctcaggtgctgcctggcaatggcaccggggacag
 gcaactcttgctgctgcagcttcccaggggagcctgtgctcacaccac
 catgaccoccttccctctctctccctagTCACTGTATCCAGCAGATCCTG
 GAGGCTGTGCTGCCACTGCCACCAGATGGGGGTGGTGCACCGGGACCTGAA
 ggtgagcaaccaccgctgggtgtggtC/Taccggggagcccctctctctct
 ctgcacgtccctA/Ctctctcagagctggacattatcagaatcgctttaat
 ttctctgtgttaggtggaaggtgggggtgggtgagattatctcccact

} Exon 6

agaggtgaggtctctgaatctgctgactgtctcttggcgctttcctccct
 tctccctcttctctggtcctgacaagCCTGAGAATCTGTTGCTGGCCCTC
 CAAGCTCAAGGGTCCCGCAGTGAAGCTGGCAGACTTTGGCCTGGCCATAG
 AGGTGGAGGGGAGCAGCAGGCATGGTTTGGtaagtgggggtgggtggca
 gaagtgtgggggtcagcaccatctgtgtG/Tacacataagcctctggga
 acccctcctgagtggtatggctggagatgctgttgggcccagccggg

} Exon 7

GGAAGGACCCGTACGGGAAGCCTGTGGACCTGTGGGCTTGTGgtgagttc
 atcctaaggcccttttgtgcccctggctgcaccggcctctggggccc
 ctctctcttcccacactcccctcccccatgtctctctcccacaggg
 ggttgaaggttctctggtcctctT/CtccagGGTCATCCTGTACATCCTGC
 TGGTTGGGTACCCCCGTTCTGGGATGAGGACCAGCACCCTGTACCAG
 CAGATCAAAGCCGGCGCTATGATgtgagtgctcctctctctcagctc
 ccatcatcctctggtgcccttccagtgaagccagcgaagccactctc
 tagttctggcacaggggtgaaccgggggagcatgaagactgaccaggg

} Exon 9

tgatgcctagccccagcagcaatgcgaggataaatgagatcatctgcat
 gaggtcttcagggtccagcacagagctctggcacatgataaggactcaaga
 aatggtagttggtgcttcattaccatcactATTATTattattattattat
 tattattattattattattctgtgctcccctctgccttacagTGCGGAA
 ACAGGAAATTATAAAAGTGACAGAGCAGCTGATTGAAGCCATAAGCAATG
 GAGATTTGAGTCTACACgtgagtcagctgggtcattctgcatgtcct
 gcctggattgtgttcagggggtaccaatggctaactttgggggtccttg

} Exon 15

caatcattcctattttcacaggtcaccctccaagggccaatgagaagC/T
 gcccagtcctcttccctctggtctcttccccagGAAGATGTGCGAC
 CCTGGCATGACAGCCTTCGAACCTGAGGCCCTGGGGAACCTGGTTGAGGG
 CCTGGACTTCCATCGATTCTATTTGAAAACCgtgagcaattccttctct
 ctgtggtgctgtgtctctctctgctccatgtcctgctggatggcacag
 caggggtacggggcagggccccagggcggtcaggagccagcgaagagga
 agctcccgtgtggactggggttggtaccctctggttccctctcctaacc

} Exon 16

gcagggcctccattgagagggctgctgggagtgagaaaagtgaagaaa
 gtgagacaggagG/Accccggttgcatctctctgcccctccatcccctc
 ctccagTGTGGTCCCGGAACAGCAAGCCGTGCACACCACCTCCTGAAT
 CCCCACATCCACCTGATGGGCGACGAGTCAGCCTGCATCGCCTACATCCG
 CATCAGCAGTACCTGGACGCTGGCGGCATCCACGc/tACCGCCAGTCCG
 AGGAGACCCGTGTCTGGCACCGCCGGGATGGCAAATGGCAGATCGTCCAC
 TTCCACAGATCTGGGGCGCCCTCCGTCTGCCCAgtaagaatcccctct
 tctctcctcccagggtaatgtaaaggatgggagtggttgcctggagggg

} Exon 17

SNP No.	Polymorphism ID	Alternative ID	Gene position
SNP1	g.33228C>T	CaM-6-SNP1	26bp downstream Exon 6
SNP2	g.33264A>C	CaM-6-SNP2	62bp downstream Exon 6
SNP3	g.36382G>T	CaM-7-SNP1	51bp downstream Exon 7
SNP4	g.37924T>C	CaM-9-SNP1	5bp upstream Exon 9
SNP5	g.58331_58336delATTATT	CaM-15-SNP1	63bp upstream Exon 15
SNP6	g.61450C>T	CaM-16-SNP1	37bp upstream Exon 16
SNP7	g.66513G>A	CaM-17-SNP1	43bp upstream Exon 17
SNP8	g.66687C>T	CaM-17-SNP2	Exon 17
SNP9	g.67057A>G	CaM-3'UTR-SNP1	3bp upstream 3'UTR
SNP10	g.67113_67114delGT	CaM-3'UTR-SNP2	3'UTR
SNP11	g.67404_67405insT	CaM-3'UTR-SNP3	3'UTR
SNP12	g.680727T>C	CaM-3'UTR-SNP4	3'UTR
SNP13	g.68818C>T	CaM-3'UTR-SNP5	3'UTR
SNP14	g.68889A>G	CaM-3'UTR-SNP6	3'UTR

Table No. 33: Polymorphisms detected by DHPLC and characterised by DNA sequencing.

5.2.8.3 Genotyping

Genotyping was performed on DNA pools, using the SNaPshot assay (see General Materials and Methods). Extension primers for the SNaPshot assay were designed for all but the Exon 15 insertion/deletion change (g.58331_58336delATTATT) since the nature of the change and its surrounding sequence did not allow for it.

SNP No	Extension primer ID	Extension primer sequence	dG (kCal/mol)
SNP1	CaM-6-SNP1-er	ggagagaggggctcccgggt	-2.1
SNP2	CaM-6-SNP2-ef	cctctctctctgcacgtccta	0
SNP3	CaM-7-SNP1-er	ggtcccagaggcttatgtgt	0
SNP4	CaM-9-SNP1-ef	gggtgaagggtctctggtcctct	-0.4
SNP6	CaM-16-SNP1-er	cagaggggaagagggactggggc	-0.7
SNP7	CaM-17-SNP1-ef	gtgaagaaagtgagacaggag	0
SNP8	CaM-17-SNP2-er	acgggtctctccgactgggcggt	-1.9
SNP9	CaM-3'UTR-SNP1-ef	aagcaccocgtgtgtgttt	-1.1
SNP10	CaM-3'UTR-SNP2-ef	ttgctgtgccgcagagatccactct	-1.7
SNP11	CaM-3'UTR-SNP3-ef	tgtaggctagaatcccgggctg	-1.9
SNP12	CaM-3'UTR-SNP4-ef	ctgtggttattggggccaagg	-0.9
SNP13	CaM-3'UTR-SNP5-er	caaccaacaccaggagggtc	-3.0
SNP14	CaM-3'UTR-SNP6-ef	agattttctgagctctccc	0

Table No. 34: Extension primer IDs, sequences and corresponding dG values

SNaPshot assays for characterised SNPs were optimised using heterozygous individuals previously identified by sequencing. The results of pool genotyping using the assay are presented in Table No. 35.

Estimated allele counts and p-values for high and average *g* pools were obtained for four SNPs (SNP2, SNP3, SNP4 and SNP7). For SNP6, SNP8 and SNP10 only one allele could be visualised in the pools, probably due to inability of the pooling method to detect low frequency alleles. For the remaining SNPs (SNP1, SNP9, SNP11, SNP12, SNP13 and SNP14), the assay failed in the pools. Since individuals heterozygous for those SNPs generated optimum height peaks for both expected alleles, it is possible that the failures in pools were due to DNA degradation as these were rather old (4 years) at

the time of the experimental analysis. Unfortunately, no further access to individual DNA samples meant that fresh pools could not be constructed and experiments repeated.

SNP No.	SNP ID	Pool ID	Estimated Allele1 count (%)	Estimated Allele2 count (%)	Allelic p-value
SNP 1	g.33228C>T		Failed		
SNP 2	g.33264A>C	mid <i>g</i> high <i>g</i>	97 (0.48) 92 (0.45)	105 (0.52) 110 (0.55)	$\chi^2=0.2485$ $p=0.618$
SNP 3	g.36382G>T	mid <i>g</i> high <i>g</i>	40 (0.20) 37 (0.18)	162 (0.80) 165 (0.82)	$\chi^2=0.1444$ $p=0.704$
SNP 4	g.37924T>C	mid <i>g</i> high <i>g</i>	96 (0.47) 100 (0.49)	106 (0.53) 102 (0.51)	$\chi^2=0.1585$ $p=0.691$
SNP 6	g.61450C>T	mid <i>g</i> high <i>g</i>	Low frequency polymorphism		
SNP 7	g.66513G>A	mid <i>g</i> high <i>g</i>	94 (0.46) 96 (0.47)	108 (0.54) 106 (0.53)	$\chi^2=0.0397$ $p=0.842$
SNP 8	g.66687C>T	mid <i>g</i> high <i>g</i>	Low frequency polymorphism		
SNP 9	g.67057A>G		Failed		
SNP 10	g.67113_67114delGT	mid <i>g</i> high <i>g</i>	Low frequency polymorphism		
SNP 11	g.67404-67405insT		Failed		
SNP 12	g.68072T>C		Failed		
SNP 13	g.68818C>T		Failed		
SNP 14	g.68889A>G		Failed		

Table No. 35: SNaPshot assay pool genotyping results with estimated allele counts, frequencies and corresponding allelic p-values. For SNPs 1, 9, 11, 12, 13 and 14 assays failed with pooled DNA, while for SNPs 6, 8 and 10 rare allele frequency was too low for allele counts to be accurately estimated with the pooling method.

None of the successfully genotyped SNPs generated a significant p-value (g.33264A>C, p=0.618; g.36382G>T, p=0.704; g.37924T>C, p=0.691 and g.66513G>A, p=0.842), as calculated from the estimated allele counts, and entered the final stage of individual genotyping.

5.2.9 Discussion

Learning and memory are elementary constituents of intelligence. These cognitive abilities and their spatial component in particular, are among the most heritable human behavioural traits. Sustained changes in synaptic strength, such as long term potentiation (LTP), are believed to be the neural basis of learning and memory ²⁶⁵. Genetic and biological investigations of the calcium/calmodulin dependent protein kinase II alpha (CaMKII- α) support the hypothesis that CaMKII- α is crucial for induction of LTP and formation of spatial memories in the hippocampus ²⁹¹. This study investigated CaMKII- α gene as candidate gene for *g*.

Using DHPLC for mutation detection fourteen intra-genic sequence changes were detected, only one of which was previously reported and deposited into the dbSNP database (g.68889A>G; rs2163766). Polymorphisms were genotyped on DNA pools by a highly accurate protocol for SNP allele frequency estimation based upon the SNaPshot primer extension chemistry ³¹³. For six polymorphisms assays optimised on individual DNA samples repeatedly failed to generate good quality electropherograms in pooled DNA (g.33228C>T, g.67057A>G, g.67404_67405insT, g.680727T>C, g.68818C>T and g.68889A>G). For those changes allele frequencies could be calculated. For g.61450C>T, g.66687C>T and g.67113_67114delGT allele frequency estimates were not obtained as only single dye electropherograms could be visualised on pooled DNA

samples. This is probably due to presence of low frequency alleles ($f < 0.05$)³¹³, rather than failed assays, as individuals heterozygous for these SNPs generated electropherograms with peaks of expected dye and height representing both alleles. Estimated allele frequencies for four SNPs at positions g.33264A>C, g.36382G>T, g.37924T>C and g.66513G>A were not significantly different between the high *g* and average *g* groups ($p_{g.33264A>C}=0.693$, $p_{g.36382G>T}=0.806$, $p_{g.37924T>C}=0.754$ and $p_{g.66513G>A}=0.910$).

Particularly interesting from the point of sequence polymorphism detection was exon11. This exon harbours the codons for the Thr²⁸⁶, Phe²⁹³ and Asn²⁹⁴ residues showed to be critical for the autophosphorylation and high-affinity calmodulin binding of the CaMKII- α protein²⁸⁵ changes in which result in LTP and learning and memory deficits²⁹⁰. The high functional importance of the protein domain this CaMKII- α exon is coding for is well reflected by a 100% amino-acid sequence match between human, rat and mouse as well as by this study's failure to detect any sequence changes in it. Such finding is within the study's expectations, as changes in the protein sequence leading to major alteration of its function are not likely to explain normal variation in human cognitive ability in the general population.

The 3' untranslated region of the CaMKII- α gene with its cis-acting regulatory elements is known to play important role in dendritic targeting of the transcript^{305,306} disruption of which results in profound spatial learning and memory impairments³⁰⁹. Thorough screening of the 3'UTR resulted in characterisation of six polymorphisms, one of which (g.67113_67114delGT) appeared to be within the putative cis-regulatory element (CNDLE). The results of pooled genotyping suggested that the deletion exists at a very low frequency in a general population (only one allele visualised in DNA pools). Bearing in mind the importance of this regulatory element, it would be of great interest to individually genotype all the subjects included in the study and to extend the research into analysis of a functional importance of the detected deletion. It is not beyond the possibility that subtle changes in the CNDLE element sequence, rather than

its complete exclusion, may result in lowered, but not abolished, efficacy of the transcript targeting to dendritic processes and subsequent reduced representation of the kinase in postsynaptic densities.

The CaMKII- α promoter region is known to be responsible for forebrain-specific expression of the gene. Mutation screening of the first 500 bp upstream from the transcription start site failed to detect any sequence changes. However, the study cannot firmly conclude that the human CaMKII- α transcription regulatory region is free from sequence polymorphisms that may influence its expression. The main reason for that is the lack of evidence for the presence of any known transcription regulatory elements in the first 500bp 5' from the transcription initiation site. The study explored two avenues in attempt to localise a putative promoter, literature and bio-informatics research. Even though there is a plethora of publications on CaMKII- α , its structure and function, a very limited amount of scientific work has been dedicated to analysis of CaMKII- α promoter. In fact, only one group reported their work on functional identification of the rat core promoter for the CaMKII- α (consensus TATA sequence at -162)³¹⁴ which influenced the decision to analyse the first few hundred base pairs in the 5' flanking region. As yet, no such work has been carried out/published on the human gene variant.

Bio-informatics findings were even more disappointing as a number of different transcription element and gene prediction programs (NIX-Nucleotide Identification of unknown sequences, <http://www.hgmp.mrc.ac.uk/>; CISTER³¹², <http://zlab.bu.edu/~mfrith/cister.shtml>) failed to recognise transcription regulatory elements of any kind not only in the first 500bp, but in the 10kb long stretch of genomic DNA upstream from the transcription start site. Given the importance of the CaMKII- α promoter it would be of great scientific interest to thoroughly examine the region immediately upstream from the TSS and identify a core sequence that is necessary and sufficient to drive the human CaMKII- α forebrain expression.

A number of polymorphisms detected in the gene were not successfully analysed on the study sample mainly for the lack of time and funding for the study to be brought to completion. These should become a priority if the study of the CaMKII- α is to be continued. One interesting avenue to explore could be stratification of the sample into those that performed particularly well in tests that examine spatial ability component of the general intelligence and those that performed less well on them, not forgetting that increasing the total sample size to at least several hundred subjects may further increase a chance of detecting smaller signals contributing to variance in g .

To summarise, sequence variants detected and analysed in the human variant of the CaMKII- α gene are not showing evidence of association with g in the studied sample and cannot explain the difference in general intelligence scores observed between the high g and average g groups.

6 GENERAL DISCUSSION

For a very long time the human mind was considered a perfect, immaterial and universal attribute of humans ¹⁰. The idea of evolution, brought about by Darwin in the 19th century, resulted in formal recognition that mental ability is a product of evolutionary processes involving natural selection and hereditary variation. However, it was Sir Francis Galton's pioneering work on the heritability of human intelligence that instigated a series of studies that resulted in important advances in data analysis, mental ability test development and the invention of factor analysis by Spearman, all of which ultimately led to the discovery of general cognitive ability, *g*.

At present, it is widely accepted that variation in general cognitive ability is significantly associated with a difference in genetic composition. The belief is firmly supported by a large amount of empirical evidence from a number of family studies carried so far. Unfortunately, suggested heritability levels range from 0.30 to 0.75, a window difficult to narrow any further with currently available data and mathematical models. The pattern of correlation in scores on mental ability tests between various relative pairs suggests complex inheritance and involvement of a number of genetic and environmental factors in shaping heritability. The most likely scenario involves a large number of genetic loci with varied but very small effect on a trait. Statistical models predict that only a small fraction of them are likely to have a modest effect and explain much of the phenotypic variance observed.

Although the heritable nature of intelligence was established a long time ago, the magnitude of this effect is still not fully recognized. Moreover, the difficulties of undertaking large, genome-wide association studies, which are currently the only way to systematically detect multiple small-effect genetic loci ¹⁰⁹, has meant that there have been relatively few molecular genetic studies of *g*. Candidate gene studies are easier to perform and indeed several have been reported in recent years ³¹⁵⁻³¹⁹. Unfortunately, the number of potential candidates is large and unlikely to be much reduced by animal

research that opened new avenues of thinking and uncovered complex interacting biological networks underlying different cognitive processes like learning and memory.

Recent advances in laboratory techniques (DNA pooling), equipment (ABI sequencers, dHPLC machines) and publicly available resources (large databases of multi-allelic and bi-allelic markers) have eliminated most of the problems encountered in the past in genome-wide studies. With these advances, it has been possible to carry out a genome-wide screen for association with IQ – the first study of this kind.

Almost 2000 micro-satellite markers with average spacing of 2 cM were tested on a case/control set, using a multiple-stage study design combined with pooled genotyping. Such a design allowed time- and cost-effective screening of a large number of markers, reduced generation of false positives to a minimum and, by incorporating a replication stage, excluded the need for p-value correction for multiple testing ¹¹⁵. However, a single marker (D4S2460), found to be associated with IQ in our case/control study, failed to be replicated by TDT testing on trios. Even though the study failed to generate markers convincingly associated with human intelligence, it showed that the DNA pooling is a reliable method and a tool of choice for future studies of this kind. It has also highlighted the need for relaxation of what appeared to be the over-stringent criteria imposed by the multiple-stage design in order to reduce the level of false negatives. By shifting the focus of the study from children to adults, known to have higher heritability levels for g , it may be possible to further improve the prospects for detecting a genuine association.

A new generation of markers, bi-allelic SNPs, is increasingly used in genetic studies. Their generous availability, dense distribution and amenability for high-throughput projects have made them the markers of choice for genome-wide studies. The question of optimal number of SNPs to be genotyped in such studies is still to be answered and suggestions are currently ranging from $\sim 250,000$ (few SNPs per 1 LDU) ¹⁴⁶ to more pessimistic $\sim 1,000,000$ SNPs (1 SNP every 3kb) ¹¹⁷. Even so, such strategy is certainly to be considered in future attempts to identify genes contributing to variation in g .

Candidate gene studies using knockout strains of mice have pointed to a number of genes having a possible role in different aspects of mental functioning. Molecular genetic analysis of two such genes, ApoE and CaMKII- α , was an attempt to throw some light on potential biological systems underlying general cognitive ability. Neither Apolipoprotein E, a well-known protein with established role in neuronal maintenance and repair^{225, 239}, nor CaMKII- α , believed to be a central molecular mechanism involved in learning and memory²⁸⁵, have shown significant association with general cognitive ability. The negative reports on these two genes do not exclude them from involvement in *g*; they rather emphasise the need for much larger study samples if small effect genes are to be detected.

Similar to combing human genome with a dense battery of genetic markers, a more systematic approach to candidate gene studies could potentially lead to significant discoveries. Rather than analysing isolated genes, investigation of whole pathways should take place. Genetic studies of such kind should run in parallel to, and not isolated from, animal research which creates an inexhaustible pool of biological hypotheses through creation of transgenic animals. CaMKII- α is a good candidate gene, but it is only one of the many components of a pathway leading to synaptic plasticity. Not only that, but its function is dependent upon other molecular interactions preceding activation of the pathway it belongs to. In addition, this pathway's activity initiates a set of cell signalling cascades crucial for memory consolidation. Clearly, investigating whole pathways and their networks is a future of candidate gene studies, but expertise in different scientific fields and joint effort of many research groups is going to be necessary to carry out and interpret results of such large studies.

The identification of genes for human intelligence is still in the initial stage, but soon the first genuine associations will start emerging. It is reasonable to believe that reports on such genes will give a new impetus to ethical criticism of the intelligence research and instigate general public concern about the wider application of the scientific knowledge. However, the criticism thus far has largely been based on

misconceptions about behavioural genetics. Therefore, it must be stressed that a large number of genes is likely to be involved in shaping human intelligence in concert with environmental factors, which makes development of genetic tests of intelligence improbable. What the critics of intelligence research often fail to appreciate is the significant contribution it provides to better understanding of molecular and cellular mechanisms of human behaviour and the mind.

As scientists, we have a responsibility to strive to further the knowledge and understanding of our environment and ourselves. In doing so, we have to make sure that the power of knowledge be not abused but that it be used solely for the enrichment of human life and the advancement of well-being of our species as a whole.

7 APPENDIX

7.1 *Genome-wide screen for association with g:*

Table with listed microsatellite markers, their positions on the Marshfield Genetic map and inter-marker distances (cM). For each of the chromosomes investigated **average** and **maximum** inter-marker distances are presented (printed in **RED** at the bottom of each chromosome's marker list).

chr1	cM	inter-marker distance (cM)	chr2	cM	inter-marker distance (cM)	chr3	cM	inter-marker distance (cM)
D1S468	4.22		D2S2584	1.95		D3S1270	6.96	
D1S2795	11.87	7.65	D2S2268	1.95	0	D3S1307	7.12	0.16
D1S214	14.04	2.17	D2S2245	7.05	5.1	D3S2426	7.38	0.26
D1S2642	14.59	0.55	D2S319	7.6	0.55	D3S1297	8.31	0.93
D1S253	14.59	0	D2S304	10.04	2.44	D3S3630	10.7	2.39
D1S1646	14.59	0	D2S2211	15.61	5.57	D3S1620	14.46	3.76
D1S548	15.13	0.54	D2S2164	18.42	2.81	D3S1515	18.38	3.92
D1S2694	15.13	0	D2S162	20.03	1.61	D3S3706	18.97	0.59
D1S1160	20.61	5.48	D2S2243	20.03	0	D3S1304	22.33	3.36
D1S503	20.61	0	D2S423	22.1	2.07	D3S3728	24.29	1.96
D1S450	20.61	0	D2S398	23.57	1.47	D3S3591	24.89	0.6
D1S2736	20.61	0	D2S2278	26.52	2.95	D3S1537	27.72	2.83
D1S1597	29.93	9.32	D2S2199	31.05	4.53	D3S1597	29.92	2.2
D1S489	29.93	0	D2S219	31.2	0.15	D3S3601	31.13	1.21
D1S2718	29.93	0	D2S2267	33.31	2.11	D3S1263	36.1	4.97
D1S2834	31.02	1.09	D2S149	34.04	0.73	D3S3680	36.1	0
D1S2728	33.75	2.73	D2S312	34.78	0.74	D3S3610	37.2	1.1
D1S407	33.75	0	D2S2346	36.72	1.94	D3S2403	37.2	0
D1S507	33.75	0	D2S305	38.87	2.15	D3S1585	38.83	1.63
D1S2697	37.05	3.3	D2S2342	40.47	1.6	D3S2338	42.1	3.27
D1S1592	38.51	1.46	D2S2373	42.65	2.18	D3S3726	42.64	0.54
D1S2826	41.92	3.41	D2S218	44.09	1.44	D3S3038	44.81	2.17
D1S199	45.33	3.41	D2S2226	45.3	1.21	D3S3659	47.44	2.63
D1S552	45.33	0	D2S2350	46.37	1.07	D3S2336	49.18	1.74
D1S2647	45.33	0	D2S165	47.43	1.06	D3S2337	50.25	1.07
D1S478	48.53	3.2	D2S2283	50.65	3.22	D3S1266	52.6	2.35
D1S458	52.7	4.17	D2S265	54.96	4.31	D3S1283	52.6	0
D1S2698	52.7	0	D2S367	54.96	0	D3S3547	55.11	2.51
D1S2838	52.7	0	D2S1325	54.96	0	D3S1609	55.11	0
D1S234	55.1	2.4	D2S1788	55.51	0.55	D3S3567	56.69	1.58
D1S2787	56.48	1.38	D2S2230	56.15	0.64	D3S1619	60.98	4.29
D1S2639	56.74	0.26	D2S2186	57.22	1.07	D3S1612	61.52	0.54
D1S2854	56.74	0	D2S2220	59.36	2.14	D3S3718	61.52	0
D1S441	64.38	7.64	D2S2163	59.36	0	D3S1260	63.12	1.6
D1S255	65.47	1.09	D2S2305	61.66	2.3	D3S3658	65.26	2.14
D1S2723	65.47	0	D2S2294	64.84	3.18	D3S2304	67.94	2.68
D1S380	67.22	1.75	D2S2174	67.58	2.74	D3S3678	68.47	0.53
D1S1598	70.41	3.19	D2S2291	69.77	2.19	D3S3647	68.47	0
D1S447	73.81	3.4	D2S2182	70.31	0.54	D3S3582	69.19	0.72
D1S451	75.66	1.85	D2S288	70.63	0.32	D3S1767	69.9	0.71
D1S2797	75.66	0	D2S123	73.61	2.98	D3S1621	70.61	0.71
D1S232	76.27	0.61	D2S2251	76.34	2.73	D3S3660	72.21	1.6
D1S2661	78.25	1.98	D2S2292	76.34	0	D3S3717	74.35	2.14
D1S475	82.41	4.16	D2S2279	77.97	1.63	D3S1592	77.01	2.66
D1S2690	83.07	0.66	D2S2165	80.69	2.72	D3S1547	78.64	1.63
D1S2867	85.68	2.61	D2S2225	82.82	2.13	D3S1313	78.64	0
D1S2890	85.68	0	D2S296	83.88	1.06	D3S1300	80.32	1.68
D1S405	86.77	1.09	D2S290	84.42	0.54	D3S3631	82.24	1.92
D1S2869	86.77	0	D2S285	86.02	1.6	D3S1312	82.24	0
D1S2741	87.86	1.09	D2S2152	87.62	1.6	D3S1600	85.97	3.73
D1S3728	89.49	1.63	D2S303	88.15	0.53	D3S1285	91.18	5.21
D1S2788	93.86	4.37	D2S2115	88.15	0	D3S3545	96.12	4.94
D1S2638	96.04	2.18	D2S2111	90.29	2.14	D3S1296	97.75	1.63
D1S515	97.49	1.45	D2S2112	90.29	0	D3S3614	101.55	3.8
D1S368	100.39	2.9	D2S2110	90.82	0.53	D3S2406	102.64	1.09
D1S1665	102.02	1.63	D2S1394	90.82	0	D3S3581	104.83	2.19
D1S2614	104.79	2.77	D2S286	94.05	3.23	D3S3653	107.19	2.36
D1S500	107.56	2.77	D2S2116	95.67	1.62	D3S3681	109.22	2.03
D1S430	109.04	1.48	D2S169	99.41	3.74	D3S1604	109.22	0
D1S2856	113.69	4.65	D2S139	101.56	2.15	D3S3049	109.22	0

chr1	cM	inter- marker distance (cM)	chr2	cM	inter- marker distance (cM)	chr3	cM	inter- marker distance (cM)
D1S454	113.69	0	D2S394	103.16	1.6	D3S1577	109.22	0
D1S551	113.69	0	D2S2161	105	1.84	D3S1276	111.89	2.67
D1S488	114.24	0.55	D2S2232	107.46	2.46	D3S3671	112.96	1.07
D1S2807	114.24	0	D2S2181	110	2.54	D3S1552	114.02	1.06
D1S2766	118.14	3.9	D2S2216	111.21	1.21	D3S1271	117.76	3.74
D1S1588	125.51	7.37	D2S2175	111.21	0	D3S3655	117.76	0
D1S435	125.51	0	D2S2209	112.28	1.07	D3S2459	119.09	1.33
D1S2868	126.16	0.65	D2S2264	114.42	2.14	D3S3632	120.43	1.34
D1S424	126.16	0	D2S2972	114.42	0	D3S1291	121.67	1.24
D1S236	128.73	2.57	D2S2356	115.49	1.07	D3S1616	124.16	2.49
D1S223	134.2	5.47	D2S135	116.02	0.53	D3S3045	124.16	0
D1S2896	134.2	0	D2S2386	118.16	2.14	D3S1281	124.16	0
D1S206	134.2	0	D2S1890	119.22	1.06	D3S3044	126.83	2.67
D1S1671	134.2	0	D2S1889	120.29	1.07	D3S2496	126.83	0
D1S2626	136.34	2.14	D2S160	122.96	2.67	D3S3683	127.89	1.06
D1S1631	136.88	0.54	D2S347	131.51	8.55	D3S3670	129.73	1.84
D1S2759	136.88	0	D2S2339	132.58	1.07	D3S1579	131.83	2.1
D1S2726	144.38	7.5	D2S1273	132.58	0	D3S1558	133.93	2.1
D1S2809	144.38	0	D2S112	141.62	9.04	D3S3649	134.64	0.71
D1S2837	146.53	2.15	D2S2219	142.83	1.21	D3S3515	136.32	1.68
D1S2744	149.2	2.67	D2S1334	145.08	2.25	D3S3576	138	1.68
D1S440	150.81	1.61	D2S314	146.15	1.07	D3S3620	138	0
D1S514	152.45	1.64	D2S2376	146.15	0	D3S3552	139.12	1.12
D1S442	154.74	2.29	D2S2288	147.4	1.25	D3S1267	139.12	0
D1S498	155.89	1.15	D2S2362	150.96	3.56	D3S3569	140.19	1.07
D1S2858	159.32	3.43	D2S381	151.5	0.54	D3S3584	143.94	3.75
D1S2624	162.57	3.25	D2S2184	152.04	0.54	D3S3607	143.94	0
D1S2125	163.34	0.77	D2S2277	154.48	2.44	D3S1292	146.6	2.66
D1S1653	164.09	0.75	D2S2299	157.55	3.07	D3S3657	148.2	1.6
D1S484	169.68	5.59	D2S2360	161.26	3.71	D3S3641	149.8	1.6
D1S104	175.62	5.94	D2S142	161.26	0	D3S1590	150.64	0.84
D1S1677	175.62	0	D2S284	161.81	0.55	D3S3528	151.49	0.85
D1S2628	177.86	2.24	D2S306	164.81	3	D3S1615	151.49	0
D1S1158	179.1	1.24	D2S2370	165.06	0.25	D3S1549	151.49	0
D1S196	181.49	2.39	D2S2380	166.51	1.45	D3S3586	152.62	1.13
D1S431	182.35	0.86	D2S399	169.41	2.9	D3S1309	153.74	1.12
D1S433	184.21	1.86	D2S124	169.41	0	D3S3546	154.48	0.74
D1S210	188.85	4.64	D2S1379	169.41	0	D3S1569	158.38	3.9
D1S218	191.52	2.67	D2S2345	171.04	1.63	D3S1550	159.8	1.42
D1S2791	192.9	1.38	D2S1776	173	1.96	D3S3704	161.04	1.24
D1S2818	198.3	5.4	D2S2284	174.3	1.3	D3S1306	164.25	3.21
D1S444	200.96	2.66	D2S2381	175.91	1.61	D3S3705	165.32	1.07
D1S191	200.96	0	D2S335	175.91	0	D3S1555	165.32	0
D1S238	202.73	1.77	D2S2307	180.79	4.88	D3S1299	166.93	1.61
D1S428	204.51	1.78	D2S300	182.96	2.17	D3S1315	168.94	2.01
D1S413	212.44	7.93	D2S385	182.96	0	D3S1584	170.14	1.2
D1S2840	212.44	0	D2S2281	186.21	3.25	D3S1570	171.47	1.33
D1S2717	218.46	6.02	D2S364	186.21	0	D3S3708	173.87	2.4
D1S1725	219.92	1.46	D2S2366	186.84	0.63	D3S2415	174.94	1.07
D1S249	220.65	0.73	D2S426	190	3.16	D3S3643	176.54	1.6
D1S205	229.13	8.48	D2S2262	190	0	D3S3622	177.75	1.21
D1S217	231.11	1.98	D2S161	191.08	1.08	D3S1614	177.75	0
D1S425	231.11	0	D2S315	193.26	2.18	D3S1243	180.8	3.05
D1S237	232.81	1.7	D2S309	198.65	5.39	D3S1282	180.8	0
D1S2827	234.52	1.71	D2S2189	199.18	0.53	D3S2433	181.87	1.07
D1S2621	238.52	4	D2S2237	200.43	1.25	D3S1574	181.87	0
D1S2871	241.26	2.74	D2S369	202.92	2.49	D3S3520	183.47	1.6
D1S2763	242.34	1.08	D2S2358	203.46	0.54	D3S2425	185.4	1.93
D1S225	245.05	2.71	D2S322	206.13	2.67	D3S2421	186.68	1.28
D1S2709	247.23	2.18	D2S317	208.58	2.45	D3S2412	189	2.32

chr1	cM	inter-marker distance (cM)	chr2	cM	inter-marker distance (cM)	chr3	cM	inter-marker distance (cM)
D1S2800	252.12	4.89	D2S137	210.43	1.85	D3S3715	190.43	1.43
D1S2850	256.26	4.14	D2S2394	210.43	0	D3S3699	191.79	1.36
D1S2785	266.27	10.01	D2S2382	213.49	3.06	D3S1618	194.68	2.89
D1S184	267.51	1.24	D2S301	214.71	1.22	D3S3578	195.6	0.92
D1S304	267.51	0	D2S2210	215.78	1.07	D3S3592	198.68	3.08
D1S321	267.51	0	D2S173	215.78	0	D3S1617	198.68	0
D1S2842	273.46	5.95	D2S295	215.78	0	D3S3570	201.14	2.46
D1S1609	274.53	1.07	D2S120	218.45	2.67	D3S1602	201.14	0
D1S423	277.8	3.27	D2S377	220.59	2.14	D3S3600	203.81	2.67
D1S2836	285.75	7.95	D2S2197	222.2	1.61	D3S1580	207.73	3.92
D1S2132	?		D2S130	222.73	0.53	D3S3530	209.41	1.68
D1S168	?		D2S2228	224.33	1.6	D3S1294	210.09	0.68
D1S3471	?		D2S2308	227	2.67	D3S2747	212.61	2.52
D1S1634	?		D2S401	229.14	2.14	D3S3557	214.45	1.84
D1S2145	?		D2S2158	229.14	0	D3S1662	214.45	0
D1S519	?		D2S2185	231.27	2.13	D3S1523	217.24	2.79
D1S2129	?		D2S2213	231.81	0.54	D3S3590	217.24	0
D1S2144	?		D2S396	232.9	1.09	D3S1305	222.83	5.59
D1S3473	?		D2S427	236.7	3.8	D3S1272	224.88	2.05
D1S3468	?		D2S2344	238.33	1.63	D3S1311	224.88	0
D1S2131	?		D2S206	240.79	2.46	D3S3707	224.88	0
D1S550	?		D2S2176	242.17	1.38	D3S3550	227.57	2.69
D1S3469	?		D2S2205	243.01	0.84	D3S1448	?	
	max	10.01	D2S2202	249.22	6.21	D3S1210	?	
	ave	2.18	D2S1397	249.22	0	D3S4550	?	
			D2S2968	251.94	2.72	D3S1449	?	
			D2S2285	258.49	6.55	D3S659	?	
			D2S125	260.63	2.14	D3S1352	?	
			D2S395	261.34	0.71	D3S1217	?	
			D2S140	263.56	2.22	D3S2389	?	
			D2S2585	263.56	0	D3S1246	?	
			D2S2338	269.07	5.51	D3S1233	?	
			D2S2736	?		D3S4544	?	
			D2S2966	?			max	5.59
			D2S1335	?			ave	1.56
			D2S1387	?				
			D2S1774	?				
			D2S2969	?				
			D2S1392	?				
			D2S2970	?				
			D2S2978	?				
			D2S2973	?				
			D2S100	?				
			D2S266	?				
			D2S207	?				
				max	9.04			
				ave	1.77			

chr4	cM	inter-marker distance (cM)	chr5	cM	inter-marker distance (cM)	chr6	cM	inter-marker distance (cM)
D4S412	4.74		D5S678	1.72		D6S942	0	
D4S3023	8.24	3.5	D5S1981	1.72	0	D6S1685	9.18	9.18
D4S431	12.35	4.11	D5S2005	1.72	0	D6S1598	11.89	2.71
D4S403	25.9	13.55	D5S2849	7.77	6.05	D6S309	14.07	2.18
D4S1602	25.9	0	D5S1980	7.77	0	D6S277	14.61	0.54
D4S3048	29.14	3.24	D5S675	9.41	1.64	D6S410	15.43	0.82
D4S1567	29.68	0.54	D5S405	9.41	0	D6S940	16.84	1.41
D4S419	33.42	3.74	D5S406	11.85	2.44	D6S470	18.22	1.38
D4S2926	33.96	0.54	D5S464	14.3	2.45	D6S1034	23.23	5.01
D4S3020	35.03	1.07	D5S635	14.91	0.61	D6S1593	24.51	1.28
D4S2953	35.56	0.53	D5S676	16.72	1.81	D6S2434	25.08	0.57
D4S3017	35.56	0	D5S1957	19.02	2.3	D6S1721	25.08	0
D4S404	37.16	1.6	D5S1976	19.67	0.65	D6S1578	26.71	1.63
D4S1551	38.77	1.61	D5S2004	21.81	2.14	D6S259	27.78	1.07
D4S3044	38.77	0	D5S432	22.88	1.07	D6S1676	29.93	2.15
D4S3052	38.77	0	D5S2081	24.48	1.6	D6S289	29.93	0
D4S1091	42.74	3.97	D5S1991	26.73	2.25	D6S1584	32.62	2.69
D4S418	44.66	1.92	D5S1954	28.22	1.49	D6S285	34.23	1.61
D4S2912	47.58	2.92	D5S416	28.76	0.54	D6S1597	37.79	3.56
D4S3027	49.47	1.89	D5S1997	30.51	1.75	D6S1029	39.2	1.41
D4S3001	49.47	0	D5S268	33.04	2.53	D6S1660	40.14	0.94
D4S2955	50.53	1.06	D5S2096	33.04	0	D6S1545	42.98	2.84
D4S2995	50.53	0	D5S2031	36.25	3.21	D6S258	44.41	1.43
D4S1581	51.6	1.07	D5S502	39.46	3.21	D6S1281	44.41	0
D4S405	56.95	5.35	D5S819	41.06	1.6	D6S1666	45.5	1.09
D4S2382	56.95	0	D5S477	45.34	4.28	D6S1629	47.71	2.21
D4S2919	57.75	0.8	D5S2062	45.34	0	D6S1583	47.71	0
D4S1547	60.16	2.41	D5S651	47.09	1.75	D6S291	49.5	1.79
D4S1536	60.79	0.63	D5S426	51.99	4.9	D6S1576	50.75	1.25
D4S3002	61.42	0.63	D5S1994	51.99	0	D6S1548	52	1.25
D4S2971	61.42	0	D5S2023	53.67	1.68	D6S1680	53.81	1.81
D4S2996	63.58	2.16	D5S1490	57.3	3.63	D6S1641	53.81	0
D4S428	64.24	0.66	D5S2022	58.55	1.25	D6S2427	53.81	0
D4S1518	68.94	4.7	D5S2092	60.92	2.37	D6S1549	61.86	8.05
D4S3000	68.94	0	D5S474	63.6	2.68	D6S1017	63.28	1.42
D4S1569	71.77	2.83	D5S407	64.67	1.07	D6S400	63.28	0
D4S3004	75.2	3.43	D5S491	66.81	2.14	D6S1582	65.14	1.86
D4S1541	75.2	0	D5S2102	66.81	0	D6S282	66.37	1.23
D4S2931	78.97	3.77	D5S398	68.03	1.22	D6S1650	68.65	2.28
D4S2969	78.97	0	D5S1956	70.44	2.41	D6S1669	73.13	4.48
D4S1517	82.09	3.12	D5S2048	73.35	2.91	D6S269	73.13	0
D4S2990	83.29	1.2	D5S2072	73.35	0	D6S1960	76.62	3.49
D4S2363	87.06	3.77	D5S647	74.07	0.72	D6S466	77.78	1.16
D4S2964	88.35	1.29	D5S2019	75.89	1.82	D6S1573	77.78	0
D4S2932	90.28	1.93	D5S1351	78.31	2.42	D6S428	77.78	0
D4S2922	90.28	0	D5S2122	78.31	0	D6S1628	80.99	3.21
D4S2361	93.48	3.2	D5S424	81.95	3.64	D6S421	84.15	3.16
D4S2460	97.3	3.82	D5S672	86.26	4.31	D6S1625	89.23	5.08
D4S1544	100.06	2.76	D5S2067	92.38	6.12	D6S460	89.83	0.6
D4S3037	100.75	0.69	D5S2094	93.59	1.21	D6S1707	90.43	0.6
D4S2909	101.55	0.8	D5S1726	94.8	1.21	D6S1634	92.25	1.82
D4S1559	104.75	3.2	D5S428	95.4	0.6	D6S1601	92.85	0.6
D4S2986	105.29	0.54	D5S459	97.21	1.81	D6S1644	96.05	3.2
D4S2961	107.95	2.66	D5S401	97.82	0.61	D6S462	99.01	2.96
D4S1570	109.02	1.07	D5S1452	99.42	1.6	D6S1570	99.01	0
D4S3026	109.02	0	D5S815	101.02	1.6	D6S450	100.91	1.9
D4S411	109.02	0	D5S484	105.29	4.27	D6S458	100.91	0
D4S1571	114.67	5.65	D5S652	105.29	0	D6S424	104.08	3.17
D4S2945	116.37	1.7	D5S2079	108.07	2.78	D6S1717	107.25	3.17
D4S406	117.06	0.69	D5S433	111.97	3.9	D6S468	107.88	0.63

chr4	cM	inter-marker distance (cM)	chr5	cM	inter-marker distance (cM)	chr6	cM	inter-marker distance (cM)
D4S2392	121.61	4.55	D5S505	111.97	0	D6S1555	107.88	0
D4S191	121.61	0	D5S475	114.75	2.78	D6S434	109.19	1.31
D4S1524	126.15	4.54	D5S2051	116.98	2.23	D6S1580	112.2	3.01
D4S430	126.15	0	D5S489	123.45	6.47	D6S1546	113.03	0.83
D4S1615	128.31	2.16	D5S404	127.93	4.48	D6S1664	113.03	0
D4S2938	129.38	1.07	D5S471	129.83	1.9	D6S447	113.61	0.58
D4S1527	132.05	2.67	D5S622	130.38	0.55	D6S278	116.26	2.65
D4S175	134.74	2.69	D5S467	131.48	1.1	D6S1698	118.08	1.82
D4S397	140.64	5.9	D5S809	133.65	2.17	D6S404	118.64	0.56
D4S1565	143.84	3.2	D5S490	133.65	0	D6S287	121.97	3.33
D4S1610	145.98	2.14	D5S808	137.46	3.81	D6S1608	122.51	0.54
D4S3008	152.98	7	D5S1995	137.95	0.49	D6S1712	122.51	0
D4S1588	154.63	1.65	D5S458	139.33	1.38	D6S408	125.71	3.2
D4S1549	154.63	0	D5S479	141.27	1.94	D6S1690	126.52	0.81
D4S3016	157.99	3.36	D5S1983	141.82	0.55	D6S1620	128.93	2.41
D4S2980	159.3	1.31	D5S2116	142.92	1.1	D6S1572	129.46	0.53
D4S3033	161.04	1.74	D5S2119	144.06	1.14	D6S457	130	0.54
D4S3046	162.47	1.43	D5S402	147.49	3.43	D6S262	130	0
D4S2952	166.85	4.38	D5S638	148.63	1.14	D6S413	131.07	1.07
D4S349	167.55	0.7	D5S413	150.34	1.71	D6S1656	131.07	0
D4S2910	171.39	3.84	D5S434	150.34	0	D6S975	133.18	2.11
D4S1545	173.59	2.2	D5S640	152.62	2.28	D6S87	133.18	0
D4S2977	176.19	2.6	D5S2014	153.17	0.55	D6S270	135.47	2.29
D4S3030	177.33	1.14	D5S2077	155.92	2.75	D6S1626	136.97	1.5
D4S1529	181.64	4.31	D5S670	156.47	0.55	D6S1699	141.15	4.18
D4S2417	181.93	0.29	D5S497	157.02	0.55	D6S314	143.4	2.25
D4S1537	183.63	1.7	D5S1499	159.22	2.2	D6S1704	144.46	1.06
D4S2956	183.63	0	D5S2049	160.87	1.65	D6S1703	146.06	1.6
D4S1607	183.63	0	D5S403	162.47	1.6	D6S1637	147.13	1.07
D4S1584	184.2	0.57	D5S422	164.19	1.72	D6S1564	149.13	2
D4S3015	185.28	1.08	D5S2066	165.13	0.94	D6S1687	153.04	3.91
D4S1530	185.28	0	D5S621	167.69	2.56	D6S290	154.1	1.06
D4S2951	185.28	0	D5S2050	171.06	3.37	D6S425	155.17	1.07
D4S3041	187.64	2.36	D5S1458	172.67	1.61	D6S442	157.31	2.14
D4S2920	190.02	2.38	D5S619	172.67	0	D6S1708	157.84	0.53
D4S2943	190.02	0	D5S504	175.34	2.67	D6S1612	159.98	2.14
D4S2954	193.74	3.72	D5S625	177.06	1.72	D6S1581	164.78	4.8
D4S171	199.93	6.19	D5S614	177.92	0.86	D6S305	166.39	1.61
D4S1540	199.93	0	D5S462	178.57	0.65	D6S411	166.39	0
D4S3051	203.77	3.84	D5S2069	182.35	3.78	D6S1579	166.39	0
MSX1	?		D5S2108	182.35	0	D6S1277	173.31	6.92
D1S503	?		D5S498	184.66	2.31	D6S1719	177.88	4.57
DRD5	?		D5S2034	187.81	3.15	D6S264	179.07	1.19
D4S2614	?		D5S469	189.23	1.42	D6S297	182.11	3.04
259	?		D5S2008	190.18	0.95	D6S1697	182.11	0
D4S622	?		D5S2073	194.88	4.7	D6S1027	187.23	5.12
	max	13.55	D5S408	195.49	0.61	D6S446	189	1.77
	ave	2.01	D5S2846	?		D6S281	190.14	1.14
			D5S1468	?		D6S1279	?	
			D5S435	?		D6S2417	?	
			D5S204	?		D6S506	?	
			D5S253	?		D6S106	?	
				max	6.47	D6S1955	?	
				ave	1.83	D6S1266	?	
							max	9.18
							ave	1.78

chr7	cM	inter-marker distance (cM)	chr8	cM	inter-marker distance (cM)	chr9	cM	inter-marker distance (cM)
D7S2477	0		D8S264	0.73		D9S1779	0	
D7S2521	5.28	5.28	D8S1824	2.92	2.19	D9S1873	9.83	9.83
D7S511	7.44	2.16	D8S1806	3.83	0.91	D9S1852	14.78	4.95
D7S3056	7.44	0	D8S518	5.63	1.8	D9S1686	14.78	0
D7S2553	13.37	5.93	D8S561	8.34	2.71	D9S168	21.88	7.1
D7S513	17.74	4.37	D8S277	8.34	0	D9S269	24.07	2.19
D7S2464	17.74	0	D8S1819	9.96	1.62	D9S268	26.24	2.17
D7S2557	23.29	5.55	D8S516	17	7.04	D9S267	26.24	0
D7S507	28.74	5.45	D8S1721	17	0	D9S235	27.32	1.08
D7S2495	29.28	0.54	D8S520	20.61	3.61	D9S1782	30.61	3.29
D7S2532	29.28	0	D8S1695	21.87	1.26	D9S1839	30.61	0
D7S2551	30.9	1.62	D8S1754	27.4	5.53	GATA124D09	32.24	1.63
D7S2535	33.05	2.15	D8S549	31.73	4.33	D9S162	34.42	2.18
D7S493	34.69	1.64	LPL	39.25	7.52	D9S171	42.73	8.31
D7S2458	36.03	1.34	D8S258	41.55	2.3	D9S1679	44.28	1.55
D7S2463	38.48	2.45	D8S1116	42.85	1.3	D9S265	44.28	0
D7S2534	40.62	2.14	D8S560	43.41	0.56	D9S761	51.21	6.93
D7S2440	41.69	1.07	D8S298	43.96	0.55	D9S161	51.81	0.6
D7S516	41.69	0	D8S1733	45.41	1.45	D9S263	51.81	0
D7S2515	43.84	2.15	D8S1739	48.79	3.38	D9S248	54.5	2.69
D7S2496	47.08	3.24	D8S1725	50.05	1.26	D9S165	58.26	3.76
D7S2492	48.15	1.07	D8S1048	54.28	4.23	D9S1817	59.34	1.08
D7S2491	48.69	0.54	D8S585	54.98	0.7	D9S1859	59.87	0.53
D7S2252	50.29	1.6	D8S1477	60.34	5.36	D9S1800	64.72	4.85
D7S656	52.7	2.41	D8S535	60.87	0.53	D9S1806	66.32	1.6
D7S484	53.5	0.8	D8S540	60.87	0	D9S226	67.93	1.61
D7S497	53.5	0	D8S255	64.6	3.73	D9S284	70.33	2.4
D7S528	57.79	4.29	D8S1817	65.13	0.53	D9S175	70.33	0
D7S2507	59.39	1.6	D8S1831	67.27	2.14	D9S1860	73.03	2.7
D7S485	59.93	0.54	D8S165	69.4	2.13	D9S1807	73.03	0
D7S2524	59.93	0	D8S285	71	1.6	D9S1123	77.56	4.53
D7S2541	61.53	1.6	D8S1113	77.89	6.89	D9S1843	80.31	2.75
D7S521	62.07	0.54	D8S260	79.36	1.47	D9S781	80.31	0
D7S671	62.87	0.8	D8S510	79.94	0.58	D9S245	80.31	0
D7S2428	65.55	2.68	D8S1748	81.68	1.74	D9S167	83.41	3.1
D7S2528	67.43	1.88	D8S512	81.68	0	D9S776	87.29	3.88
D7S2436	67.96	0.53	D8S1840	82.84	1.16	D9S1865	87.29	0
D7S1508	69.56	1.6	D8S1767	83.51	0.67	D9S1680	90.4	3.11
D7S2467	73.31	3.75	D8S543	87.54	4.03	D9S1812	90.4	0
D7S2542	74.91	1.6	D8S1807	90.33	2.79	D9S283	94.85	4.45
D7S659	75.98	1.07	D8S1776	92.58	2.25	D9S1820	95.65	0.8
D7S2549	77.91	1.93	D8S1829	94.61	2.03	D9S1836	96.99	1.34
D7S2530	77.91	0	D8S501	95.15	0.54	D9S1796	97.53	0.54
D7S2489	78.65	0.74	D8S1736	97.28	2.13	D9S1781	99.4	1.87
D7S2483	78.65	0	D8S1707	101.69	4.41	D9S196	101.27	1.87
D7S502	78.65	0	D8S273	102.62	0.93	D9S1851	103.42	2.15
D7S645	80.42	1.77	D8S1724	103.69	1.07	D9S1809	103.42	0
D7S2435	80.42	0	D8S270	103.69	0	D9S910	104.48	1.06
D7S2476	84.52	4.1	D8S1772	107.97	4.28	D9S1690	106.63	2.15
D7S1870	86.12	1.6	D8S546	113.16	5.19	D9S1866	110.91	4.28
D7S2490	87.38	1.26	D8S521	116.14	2.98	D9S172	111.99	1.08
D7S2470	89.88	2.5	D8S1834	117.62	1.48	D9S2026	117.37	5.38
D7S2499	91.67	1.79	D8S1714	117.62	0	D9S261	117.37	0
D7S634	93.1	1.43	D8S1784	118.15	0.53	D9S105	120.04	2.67
D7S660	93.63	0.53	D8S200	120.29	2.14	D9S930	120.04	0
D7S2212	95.43	1.8	D8S539	123.54	3.25	D9S1683	120.04	0
D7S2417	97.38	1.95	D8S1015	123.54	0	D9S51	123.33	3.29
D7S524	97.38	0	D8S555	123.54	0	D9S1776	123.33	0
D7S630	98.44	1.06	D8S1694	125.27	1.73	D9S1811	126.41	3.08
D7S492	100.05	1.61	D8S1726	130	4.73	D9S754	126.41	0

chr10	cM	inter-marker distance (cM)	chr11	cM	inter-marker distance (cM)	chr12	cM	inter-marker distance (cM)
D10S249	2.13		D11S1758	7.03		D12S352	0	
D10S1706	5.21	3.08	D11S2362	8.9	1.87	D12S91	1.71	1.71
D10S591	13.49	8.28	D11S1760	11.67	2.77	D12S1656	2.79	1.08
D10S552	14.33	0.84	D11S1331	12.92	1.25	D12S100	3.33	0.54
D10S189	19	4.67	D11S1338	12.92	0	D12S1689	3.33	0
D10S547	29.15	10.15	D11S926	20.4	7.48	D12S1615	4.64	1.31
D10S585	30	0.85	D11S2368	22.56	2.16	D12S372	6.42	1.78
D10S527	33.48	3.48	D11S928	26.34	3.78	D12S1626	7.12	0.7
D10S1721	33.48	0	D11S1359	29.18	2.84	D12S1725	9.52	2.4
D10S1725	36.06	2.58	D11S1750	33.02	3.84	D12S314	11.37	1.85
D10S1653	40.36	4.3	D11S904	33.57	0.55	D12S99	12.6	1.23
D10S1476	42.5	2.14	D11S1392	43.16	9.59	D12S356	14.23	1.63
D10S211	48.36	5.86	D11S935	45.94	2.78	D12S336	19.68	5.45
D10S197	52.1	3.74	D11S4185	45.94	0	D12S1695	19.68	0
D10S572	52.1	0	D11S1279	50.88	4.94	D12S1690	20.27	0.59
D10S1641	54.23	2.13	D11S1763	53.56	2.68	D12S89	23.41	3.14
D10S588	54.23	0	D11S903	54.75	1.19	D12S98	24.45	1.04
D10S1684	57.42	3.19	D11S1777	58.4	3.65	D12S1581	29.73	5.28
D10S193	59.03	1.61	D11S1335	58.4	0	D12S364	30.6	0.87
D10S208	60.64	1.61	D11S1765	61.78	3.38	D12S308	31.22	0.62
D10S1654	62.23	1.59	D11S913	67.48	5.7	D12S1303	32.48	1.26
D10S1791	65.97	3.74	D11S987	67.48	0	D12S1728	34.85	2.37
D10S578	65.97	0	D11S4184	75.3	7.82	D12S310	36.06	1.21
D10S1787	68.63	2.66	D11S916	76.13	0.83	D12S1650	38.5	2.44
D10S225	70.06	1.43	D11S1789	79.98	3.85	D12S1682	38.5	0
D10S196	70.23	0.17	D11S937	79.98	0	D12S1654	41.55	3.05
D10S538	70.23	0	D11S901	85.48	5.5	D12S1688	42.1	0.55
D10S539	72.9	2.67	D11S4135	86.98	1.5	D12S1591	43.38	1.28
D10S546	75.57	2.67	D11S1354	87.89	0.91	D12S1617	44.03	0.65
D10S1762	75.57	0	D11S4082	89.69	1.8	D12S1596	45.95	1.92
D10S464	79.03	3.46	D11S1780	90.29	0.6	D12S1643	50.9	4.95
D10S549	79.03	0	D11S4175	91.47	1.18	D12S1584	51.99	1.09
D10S1225	80.77	1.74	D11S1332	93.12	1.65	D12S331	54.46	2.47
D10S1652	80.77	0	D11S898	98.98	5.86	D12S1592	56.38	1.92
D10S1640	80.77	0	D11S1886	100.62	1.64	D12S85	61.34	4.96
D10S1743	82.5	1.73	D11S2017	101.75	1.13	D12S1713	61.34	0
D10S1646	86.2	3.7	D11S1391	104.6	2.85	D12S1661	63.89	2.55
D10S1647	88.41	2.21	DRD2	105.17	0.57	D12S1627	64.43	0.54
D10S537	91.13	2.72	D11S4206	105.17	0	D12S1629	65.49	1.06
D10S188	93.92	2.79	D11S908	108.59	3.42	D12S368	66.03	0.54
D10S1136	97.29	3.37	D11S4129	115.53	6.94	D12S1622	69.82	3.79
D10S202	98.41	1.12	D11S925	118.47	2.94	D12S90	71.61	1.79
D10S573	106.11	7.7	D11S1353	122.47	4	D12S1632	71.61	0
D10S1735	109.33	3.22	D11S4094	123	0.53	D12S1644	72.2	0.59
D10S1753	112.58	3.25	D11S933	124.07	1.07	D12S305	74.58	2.38
D10S1755	114.19	1.61	D11S990	126.21	2.14	D12S83	75.17	0.59
D10S185	116.34	2.15	D11S4158	126.21	0	D12S1655	75.17	0
D10S1758	118.94	2.6	D11S4151	127.33	1.12	D12S1601	75.17	0
D10S1726	121.98	3.04	D11S4110	129.02	1.69	D12S1585	75.76	0.59
D10S597	128.73	6.75	D11S2367	138.56	9.54	D12S1676	78.74	2.98
D10S1693	137.39	8.66	D11S969	146.6	8.04	D12S313	79.93	1.19
D10S542	138.47	1.08	D11S968	147.77	1.17	D12S1680	80.52	0.59
D10S587	147.57	9.1	D11S4125	147.77	0	D12S92	83.19	2.67
D10S217	157.89	10.32	HBB	?		D12S80	83.19	0
D10S1676	160.04	2.15	D11S4967	?		D12S326	86.4	3.21
D10S1655	162.38	2.34	D11S2018	?		D12S1709	86.4	0
D10S1651	168.77	6.39	FCER1B	?		D12S106	90.77	4.37
D10S1675	170.4	1.63	D11S2008	?		D12S1719	94.49	3.72
D10S590	170.4	0	D11S490	?		D12S316	94.49	0
D10S212	170.94	0.54	D11S4966	?		D12S1345	96.09	1.6

chr10	cM	inter-marker distance (cM)	chr11	cM	inter-marker distance (cM)	chr12	cM	inter-marker distance (cM)
D10S179	?		D11S863	?		D12S1346	97.16	1.07
D10S1257	?		D11S2011	?		D12S348	100.92	3.76
D10S1187	?			max	9.59	D12S101	100.92	0
D10S1555	?			ave	2.71	D12S302	101.98	1.06
D10S71	?					D12S332	105.18	3.2
	max	10.32				D12S1727	107.19	2.01
	ave	2.86				D12S1607	107.86	0.67
						D12S360	111.27	3.41
						D12S338	111.87	0.6
						D12S317	114.28	2.41
						D12S353	115.18	0.9
						D12S1613	116.08	0.9
						D12S1605	116.66	0.58
						D12S1339	118.68	2.02
						D12S105	118.68	0
						D12S1616	119.55	0.87
						D12S1665	125.31	5.76
						D12S1602	125.31	0
						D12S79	125.31	0
						D12S1718	128.05	2.74
						D12S366	133.33	5.28
						D12S349	134.54	1.21
						D12S76	136.82	2.28
						D12S1666	136.82	0
						D12S395	136.82	0
						D12S321	136.82	0
						D12S1603	140.17	3.35
						D12S1614	144.83	4.66
						D12S340	146.39	1.56
						D12S324	147.17	0.78
						D12S1658	148.24	1.07
						D12S1675	150.7	2.46
						D12S1679	153.19	2.49
						D12S1609	153.33	0.14
						D12S1659	155.94	2.61
						D12S60	159.59	3.65
						D12S1714	159.59	0
						D12S1045	160.68	1.09
						D12S97	160.68	0
						D12S343	163.55	2.87
						D12S1599	164.63	1.08
						D12S1628	165.69	1.06
						D12S1638	168.79	3.1
						D12S70	?	
						D12S2194	?	
						D12S833	?	
						D12S2196	?	
						max	5.76	
						ave	1.65	

chr13	cM	inter-marker distance (cM)	chr14	cM	inter-marker distance (cM)	chr15	cM	inter-marker distance (cM)
D13S1236	2.77		D14S132	0		D15S128	6.11	
D13S175	6.03	3.26	D14S261	6.46	6.46	D15S97	9.85	3.74
D13S250	6.03	0	D14S50	12.46	6	D15S156	14.58	4.73
D13S232	6.99	0.96	D14S283	13.89	1.43	D15S1002	14.58	0
D13S292	8.87	1.88	D14S972	21.51	7.62	D15S1019	19.12	4.54
D13S1243	9.79	0.92	D14S64	22.66	1.15	D15S165	20.24	1.12
D13S283	11.45	1.66	D14S1032	23.2	0.54	D15S231	24.06	3.82
D13S1285	12.91	1.46	D14S597	28.01	4.81	D15S144	25.3	1.24
D13S1294	12.91	0	D14S1042	28.01	0	D15S1007	25.86	0.56
D13S1244	15.19	2.28	D14S615	28.01	0	D15S1040	28.35	2.49
D13S625	16.2	1.01	D14S1040	31.75	3.74	D15S971	31.46	3.11
D13S217	17.21	1.01	D14S1060	34.43	2.68	D15S118	32.58	1.12
D13S1250	17.21	0	D14S741	36.76	2.33	D15S1012	35.95	3.37
D13S1242	17.21	0	D14S70	40.11	3.35	D15S978	45.62	9.67
D13S629	19.36	2.15	D14S1049	40.94	0.83	D15S1003	47.85	2.23
D13S1299	19.36	0	D14S253	42.81	1.87	D15S962	47.85	0
D13S1246	20.44	1.08	D14S69	45.12	2.31	D15S117	51.21	3.36
D13S1287	21.51	1.07	D14S278	45.12	0	D15S987	59.05	7.84
D13S289	21.51	0	D14S266	47.51	2.39	D15S993	59.61	0.56
D13S260	23.65	2.14	D14S288	47.51	0	D15S1009	61.28	1.67
D13S171	25.08	1.43	D14S976	50.5	2.99	D15S651	61.84	0.56
D13S267	26.87	1.79	D14S1068	50.5	0	D15S213	62.4	0.56
D13S1293	26.87	0	D14S1031	52.49	1.99	D15S153	62.4	0
D13S305	28.04	1.17	D14S991	55.82	3.33	D15S1025	69.46	7.06
D13S219	28.87	0.83	D14S276	56.36	0.54	D15S650	70.73	1.27
D13S218	32.9	4.03	D14S1064	57.43	1.07	D15S215	72.94	2.21
D13S263	38.32	5.42	D14S66	59.43	2	D15S1027	74.69	1.75
D13S1247	38.96	0.64	D14S994	66.81	7.38	D15S1037	75.85	1.16
D13S1272	41.71	2.75	D14S997	67.99	1.18	D15S1041	77.69	1.84
D13S291	41.71	0	D14S63	69.18	1.19	D15S205	78.92	1.23
D13S326	42.8	1.09	D14S1069	73.03	3.85	D15S1045	85.64	6.72
D13S287	45	2.2	D14S1065	74.01	0.98	D15S127	86.81	1.17
D13S155	45.55	0.55	D14S540	74.01	0	D15S649	98.44	11.63
D13S1301	47.19	1.64	D14S1011	74.96	0.95	D15S1038	100.59	2.15
D13S233	47.19	0	D14S258	76.28	1.32	D15S130	100.59	0
D13S1320	49.38	2.19	D14S268	79.94	3.66	D15S657	104.86	4.27
D13S1231	51.57	2.19	D14S74	87.36	7.42	D15S1014	107.71	2.85
D13S275	52.1	0.53	D14S1022	92.69	5.33	D15S212	109.29	1.58
D13S1324	53.17	1.07	D14S140	95.89	3.2	D15S985	110.94	1.65
D13S1257	55.31	2.14	D14S256	96.42	0.53	D15S966	112.58	1.64
D13S152	55.31	0	D14S1015	105	8.58	D15S120	112.58	0
D13S269	56.38	1.07	D14S280	105	0	D15S514	?	
D13S792	57.46	1.08	D14S617	105.53	0.53	D15S112	?	
D13S1281	59.79	2.33	D14S1050	107.13	1.6	D15S222	?	
D13S160	61.17	1.38	D14S749	108.22	1.09	D15S172	?	
D13S921	62.83	1.66	D14S265	113.17	4.95	D15S111	?	
D13S264	64.97	2.14	D14S987	114.81	1.64		max	11.63
D13S251	64.97	0	D14S51	115.6	0.79		ave	2.66
D13S271	64.97	0	D14S979	115.91	0.31			
D13S253	66.59	1.62	D14S65	117.3	1.39			
D13S266	67.65	1.06	D14S998	117.75	0.45			
D13S265	68.73	1.08	D14S78	125.88	8.13			
D13S795	71.06	2.33	D14S1426	125.88	0			
D13S1300	71.06	0	D14S985	126.61	0.73			
D13S167	73.04	1.98	D14S293	133.03	6.42			
D13S1241	76.26	3.22	D14S118	133.03	0			
D13S1252	77.47	1.21	D14S260	134.3	1.27			
D13S1284	79.49	2.02	D14S292	134.3	0			
D13S779	82.93	3.44	D14S1007	138.18	3.88			
D13S225	83.57	0.64	D14S582	?				

chr13	cM	inter- marker distance (cM)	chr14	cM	inter- marker distance (cM)
D13S158	84.87	1.3	D14S1278	?	
D13S1256	84.87	0	D14S750	?	
D13S1323	84.87	0	D14S1433	?	
D13S1311	90.27	5.4		max	8.58
GATA135E01	90.27	0		ave	2.38
D13S286	93.52	3.25			
D13S278	93.52	0			
D13S796	93.52	0			
D13S778	96.7	3.18			
D13S1265	98.82	2.12			
D13S1315	102.73	3.91			
D13S1295	110.55	7.82			
D13S285	110.55	0			
D13S293	114.98	4.43			
D13S141	?				
D13S127	?				
D13S789	?				
D13S115	?				
D13S138	?				
D13S231	?				
D13S120	?				
D13S894	?				
D13S1818	?				
D13S802	?				
D13S1491	?				
	max	7.82			
	ave	1.54			

chr16	cM	inter-marker distance (cM)	chr17	cM	inter-marker distance (cM)	chr18	cM	inter-marker distance (cM)
D16S521	1.08		D17S1308	0.63		D18S59	0	
D16S3124	7.05	5.97	D17S831	6.60	5.97	D18S1146	2.08	2.08
D16S3065	8.16	1.11	D17S829	10.02	3.42	D18S476	2.84	0.76
D16S423	10.36	2.2	D17S1828	10.02	0	D18S481	6.94	4.1
D16S3042	13.12	2.76	D17S1298	10.72	0.7	D18S1138	8.3	1.36
D16S3092	13.67	0.55	D17S796	14.69	3.97	D18S63	8.3	0
D16S3030	13.67	0	D17S960	15.77	1.08	D18S471	18.7	10.4
D16S3108	14.77	1.1	D17S1805	16.84	1.07	D18S452	18.7	0
D16S3087	16.97	2.2	D17S954	22.24	5.4	D18S458	18.7	0
D16S687	18.07	1.1	D17S1303	23.56	1.32	D18S1163	24.08	5.38
D16S414	20.77	2.7	D17S799	31.96	8.4	D18S464	31.17	7.09
D16S519	20.77	0	D17S936	31.96	0	D18S1153	35.46	4.29
D16S2613	22.02	1.25	D17S1856	35.55	3.59	D18S1150	37.15	1.69
D16S748	22.65	0.63	D17S921	36.14	0.59	D18S1158	38.92	1.77
D16S3060	28.3	5.65	D17S1857	43.01	6.87	D18S1116	41.24	2.32
D16S764	29.97	1.67	D17S1794	47	3.99	D18S53	41.24	0
D16S3017	32.07	2.1	D17S935	50.74	3.74	D18S1149	49.55	8.31
D16S3056	35.44	3.37	D17S1863	50.74	0	D18S1107	51.21	1.66
D16S749	39.04	3.6	D17S1800	51.63	0.89	D18S478	52.86	1.65
D16S3045	40.65	1.61	D17S1293	56.48	4.85	D18S819	52.86	0
D16S417	43.89	3.24	D17S1842	57.71	1.23	D18S463	56.71	3.85
D16S420	44.45	0.56	D17S1846	57.71	0	D18S456	61.73	5.02
D16S3068	48.53	4.08	D17S1299	62.01	4.3	D18S1128	62.84	1.11
D16S3116	50.6	2.07	D17S1834	64.16	2.15	D18S1102	62.84	0
D16S769	50.6	0	D17S1868	64.16	0	D18S1130	64.48	1.64
D16S3145	52.26	1.66	D17S1795	68.44	4.28	D18S475	64.48	0
D16S540	57.79	5.53	D17S1877	74.45	6.01	D18S454	66.66	2.18
D16S536	58.46	0.67	D17S787	74.99	0.54	D18S1118	68.91	2.25
D16S409	58.46	0	D17S1799	74.99	0	D18S1110	71.32	2.41
D16S517	58.46	0	D17S1853	80.38	5.39	D18S539	74.93	3.61
D16S3080	59.68	1.22	D17S944	82.56	2.18	D18S484	74.93	0
D16S3136	62.11	2.43	D17S942	85.94	3.38	D18S69	77.36	2.43
D16S3396	63.78	1.67	D17S795	89.32	3.38	D18S1152	80.41	3.05
D16S757	65.1	1.32	D17S2059	93.27	3.95	D18S1155	83.46	3.05
D16S2623	66.1	1	D17S1350	93.98	0.71	D18S64	84.8	1.34
D16S419	67.4	1.3	D17S1351	95.99	2.01	D18S466	104.17	19.37
D16S415	67.62	0.22	D17S1823	97.6	1.61	D18S848	106.81	2.64
D16S3034	69.05	1.43	D17S1602	99.21	1.61	D18S1106	106.81	0
D16S771	70.69	1.64	D17S1839	102.46	3.25	D18S469	109.18	2.37
D16S3253	71.77	1.08	D17S1817	103.53	1.07	D18S58	111.2	2.02
D16S3039	74.44	2.67	D17S785	103.53	0	D18S1009	116.44	5.24
D16S494	77.15	2.71	UT952	105.68	2.15	D18S498	?	
D16S3143	81.15	4	D17S937	105.68	0	D18S977	?	
D16S3129	81.95	0.8	D17S1847	111.22	5.54	D18S64	?	
D16S3111	83.55	1.6	D17S836	112.92	1.7	D18S1376	?	
D16S400	83.55	0	D17S784	116.86	3.94	D18S812	?	
D16S3095	87.06	3.51	D17S1822	116.86	0	D18S378	?	
D16S512	89.63	2.57	D17S928	126.46	9.6	MBP	?	
D16S3115	92.1	2.47	D17S516	?		D18S37	?	
D16S3101	93.78	1.68	D17S71	?		D18S862	?	
D16S3049	97.03	3.25	D17S2195	?				
D16S3096	99.44	2.41						
D16S516	100.39	0.95			max		19.37	
D16S3104	105.17	4.78			ave		2.91	
D16S3119	105.17	0						
D16S3098	108.32	3.15						
D16S505	108.96	0.64						
D16S3091	111.12	2.16						
D16S422	111.12	0						
D16S3061	121.45	10.33						

chr16	cM	inter- marker distance (cM)
D16S539	124.73	3.28
D16S520	125.82	1.09
D16S3063	127.99	2.17
D16S3077	127.99	0
D16S3023	132.6	4.61
PKD1	?	
D16S310	?	
D16S292	?	
D16S301	?	
D16S291	?	
D16S3098	?	
D16S79	?	
D16S488	?	
D16S319	?	
D16S3023	?	
D16S481	?	
D16S541	?	
D16S2618	?	
D16S3255	?	
	max	10.33
	ave	2.05

chr19	cM	inter-marker distance (cM)	chr20	cM	inter-marker distance (cM)	chr21	cM	inter-marker distance (cM)
D19S886	0		D20S864	0		D21S1904	2.13	
D19S424	10.97	10.97	D20S117	2.83	2.83	D21S1431	2.13	0
D19S209	10.97	0	D20S113	8.97	6.14	D21S120	2.99	0.86
D19S216	20.01	9.04	D20S181	9.53	0.56	D21S1256	9.72	6.73
D19S427	20.75	0.74	D20S889	11.2	1.67	D21S1899	9.72	0
D19S177	20.75	0	D20S116	11.2	0	D21S1905	11.3	1.58
D19S406	25.17	4.42	D20S895	13.98	2.78	D21S1918	16.22	4.92
D19S583	32.94	7.77	D20S192	18.79	4.81	D21S210	20.45	4.23
D19S914	36.22	3.28	D20S448	18.79	0	D21S1907	20.45	0
D19S415	42.28	6.06	D20S900	21.15	2.36	D21S1896	21.88	1.43
D19S385	42.28	0	D20S851	24.7	3.55	D21S260	22.59	0.71
D19S432	42.28	0	D20S901	26.13	1.43	D21S1916	25.26	2.67
D19S593	45.48	3.2	D20S188	29.09	2.96	D21S1913	25.8	0.54
D19S925	48.52	3.04	D20S894	30.56	1.47	D21S263	27.4	1.6
D19S434	49.75	1.23	D20S163	31.43	0.87	D21S1252	35.45	8.05
D19S414	54.01	4.26	D20S186	32.3	0.87	D21S1894	35.45	0
D19S874	58.69	4.68	D20S172	35.51	3.21	D21S1917	38.65	3.2
D19S245	58.69	0	D20S904	37.65	2.14	D21S1893	43.67	5.02
D19S587	59.36	0.67	D20S98	37.65	0	D21S1887	45.87	2.2
D19S220	62.03	2.67	D20S112	39.25	1.6	D21S266	45.87	0
D19S417	63.1	1.07	D20S182	39.9	0.65	D21S1260	46.71	0.84
D19S420	66.3	3.2	D20S471	42.28	2.38	D21S1839	?	
D19S913	67.37	1.07	D20S54	45.77	3.49	D21S368	?	
D19S902	72.72	5.35	D20S912	46.71	0.94	D21S1235	?	
D19S879	75.41	2.69	D20S837	50.81	4.1	D21S1830	?	
D19S867	77.54	2.13	D20S174	54.09	3.28	D21S1264	?	
D19S585	79.48	1.94	D20S855	56.83	2.74	D21S1261	?	
D19S601	83.19	3.71	D20S99	56.83	0	D21S409	?	
D19S571	84.08	0.89	D20S899	58.48	1.65	D21S82	?	
D19S924	88.85	4.77	D20S911	61.77	3.29	pfl	?	
D19S418	92.56	3.71	D20S481	62.32	0.55	D21S366	?	
D19S887	100.01	7.45	D20S836	64.88	2.56	D21S416	?	
D19S210	100.01	0	D20S197	66.16	1.28		max	8.05
D19S544	100.01	0	D20S178	66.16	0		ave	2.23
D19S214	100.61	0.6	D20S159	69.5	3.34			
D19S254	100.61	0	D20S176	72.27	2.77			
D19S711	?		D20S185	77.75	5.48			
D19S418	?		D20S854	80.63	2.88			
D19S1171	?		D20S120	83.51	2.88			
DMD	?		D20S100	84.78	1.27			
D19S414	?		D20S102	86.98	2.2			
	max	10.97	D20S171	95.7	8.72			
	ave	2.87	D20S173	98.09	2.39			
			D20S164	101.22	3.13			
			D20S57	?				
			D20S1143	?				
			D20S1149	?				
			D20S477	?				
			D20S151	?				
			D20S110	?				
			D20S56	?				
			D20S43	?				
			max	8.72				
			ave	2.35				

chr22	cM	inter- marker distance (cM)
D22S420	4.06	
D22S539	14.44	10.38
D22S257	17.71	3.27
D22S926	21.47	3.76
D22S315	21.47	0
D22S429	22.01	0.54
D22S533	22.01	0
D22S1150	28.57	6.56
D22S1176	29.66	1.09
D22S280	31.3	1.64
D22S685	32.39	1.09
D22S422	32.93	0.54
D22S683	36.22	3.29
D22S1173	37.82	1.6
D22S1156	44.32	6.5
D22S284	46.42	2.1
D22S1171	48.19	1.77
D22S279	48.19	0
D22S1153	51.54	3.35
D22S444	51.54	0
D22S274	51.54	0
D22S1161	59.5	7.96
D22S1169	60.61	1.11
D22S431	?	
D22S304	?	
CRYBB2	?	
D22S572	?	
	max	10.38
	ave	2.57

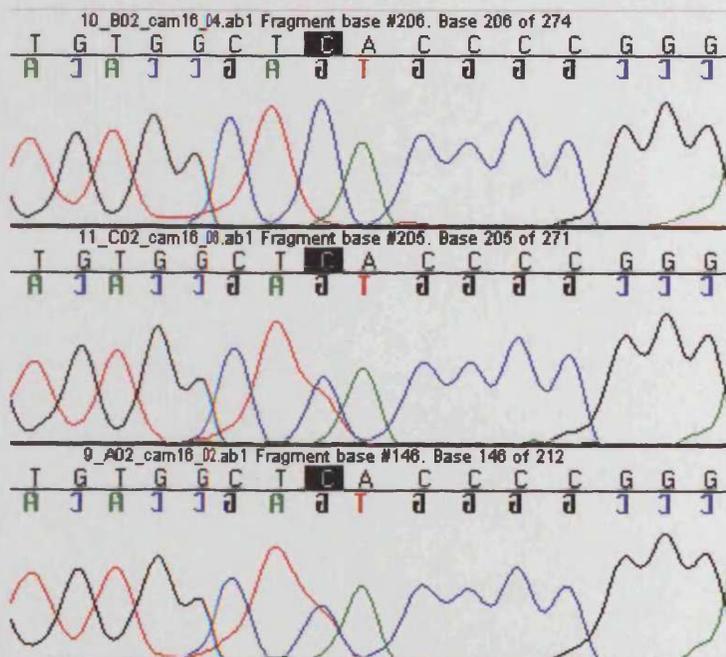
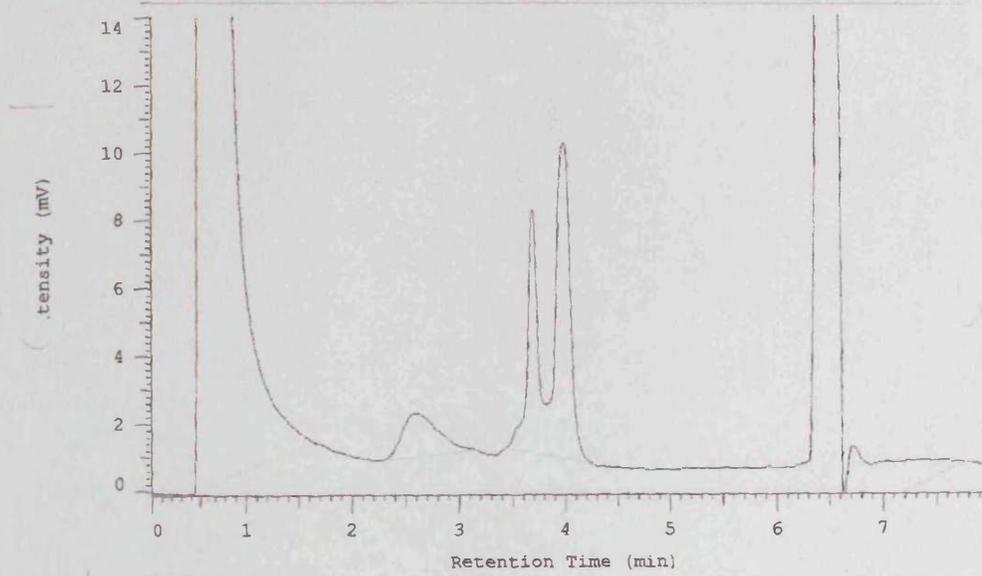
7.2 Investigating Calcium/calmodulin dependant protein kinase II alpha (CaMKII- α) as a candidate gene for g

Selected dHPLC elution profile images were generated by amplified genomic DNA fragments speculated to harbour sequence polymorphisms. Profiles are accompanied by their respective sequencing electropherograms.

CaMKII-exo 6 - SNP 1 (C/T)
Heterozygous individuals: 026R, 029R

CaMK-new-exo - Vial 15 Inj 1 015 - Channel 1

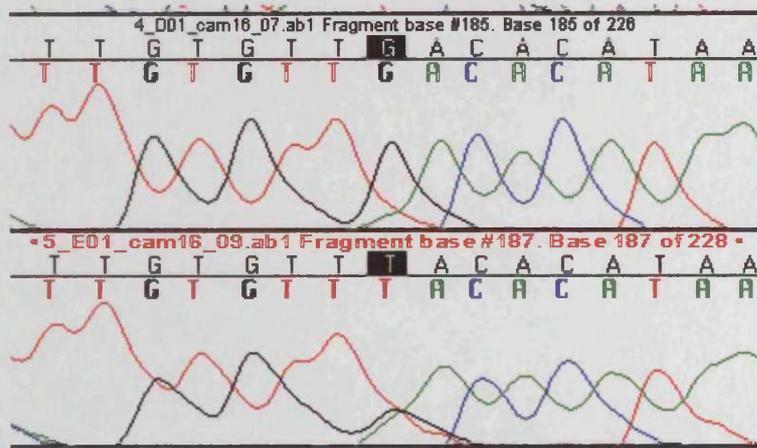
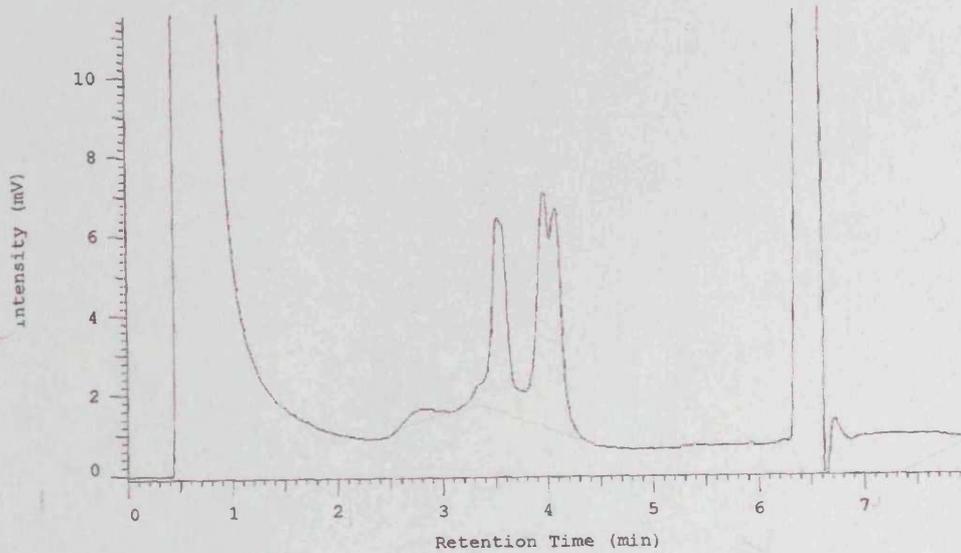
Current Data Path: D:\Dragana2\DATA\0144
Data Desc.: IFM CH1 2-D
Vial Number: 15 Inj Number: 1 Sample Name: 015



CaMKII-exo7 – SNP 1 (G/A)
Heterozygous individuals: 028R

CaMK-new-exo - Vial 30 Inj 1 030 - Channel 1

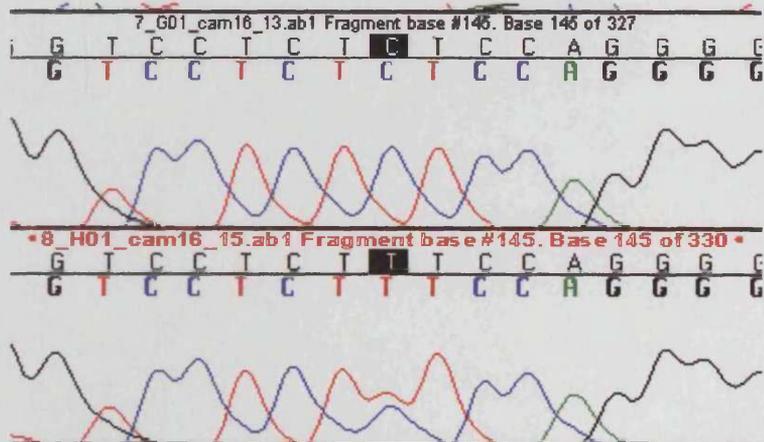
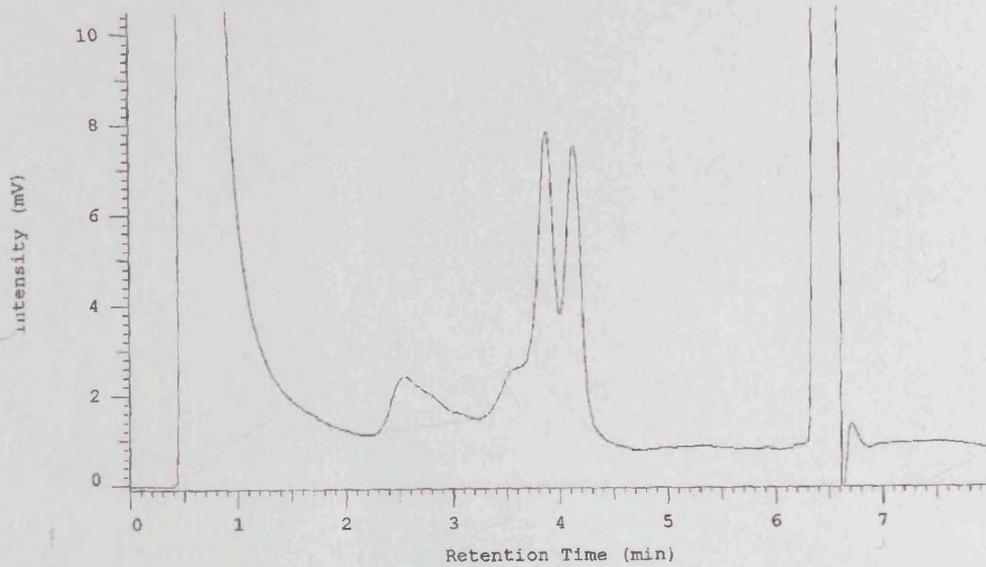
Current Data Path: D:\Dragana2\DATA\0145
Data Desc.: IFM CHI 2-D
Vial Number: 30 Inj Number: 1 Sample Name: 030



CaMKII-exo9 – SNP 1 (C/T)
Heterozygous individuals: 028R

CaMK-new-exo - Vial 47 Inj 1 047 - Channel 1

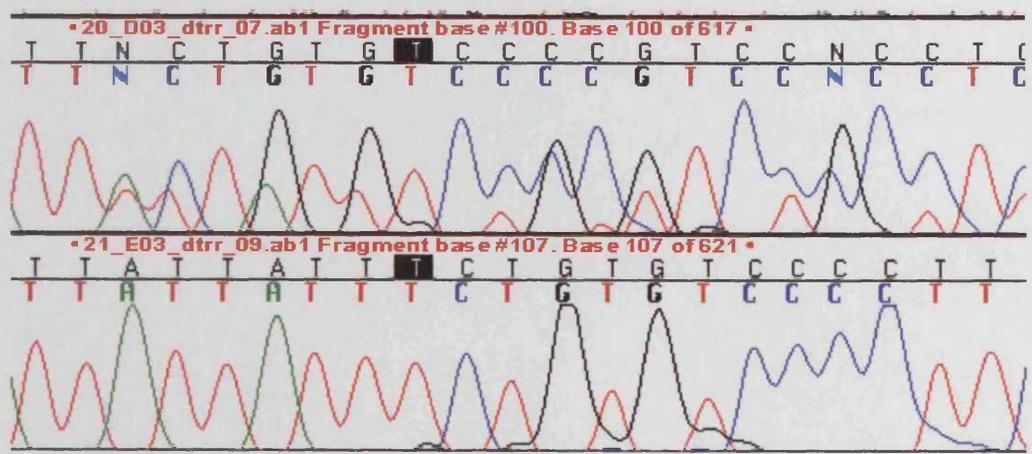
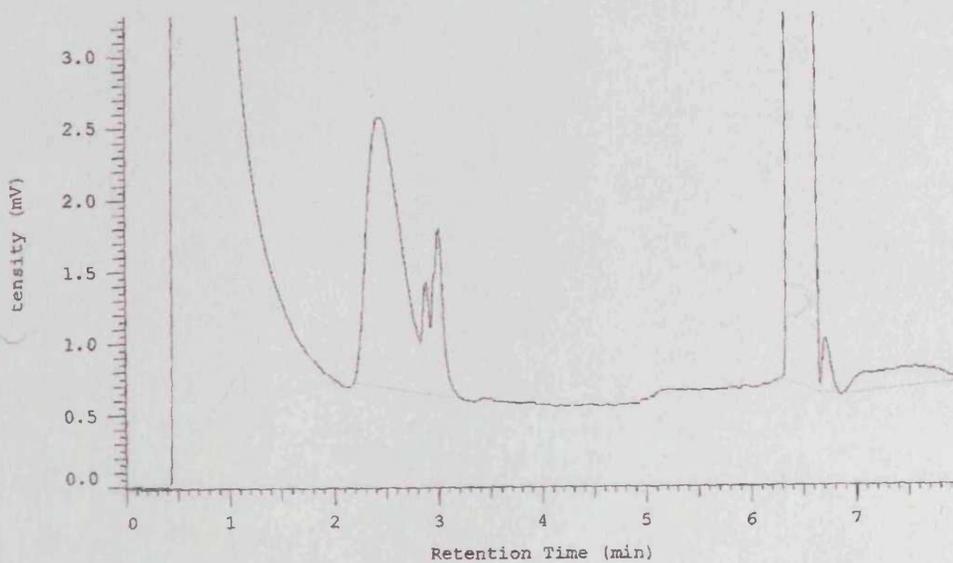
Current Data Path: D:\Dragana2\DATA\0146
Data Desc.: IFM CH1 2-D
Vial Number: 47 Inj Number: 1 Sample Name: 047



CaMKII-exo15 – SNP 1 (ATTATT deletion)
Heterozygous individuals: 020R

CaMK-new-exo - Vial 58 Inj 1 058 - Channel 1

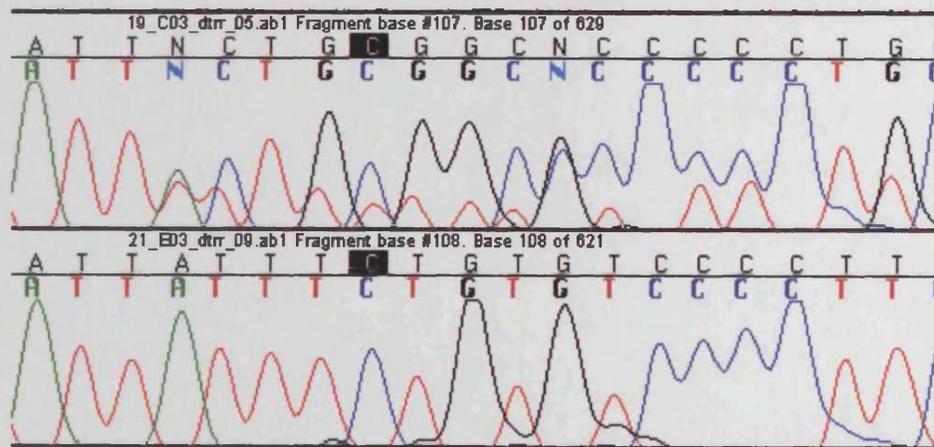
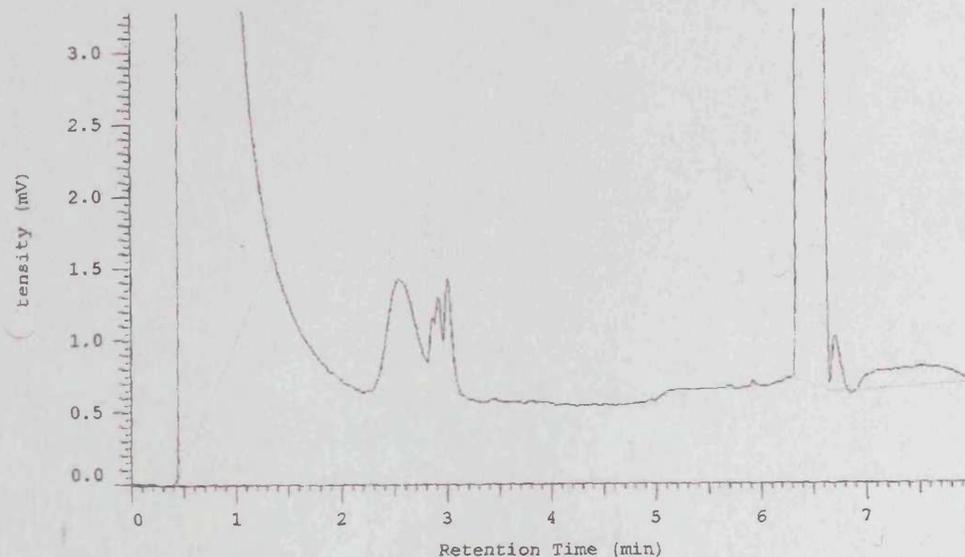
Current Data Path: D:\Dragana2\DATA\0140
Data Desc.: IFM CH1 2-D
Vial Number: 58 Inj Number: 1 Sample Name: 058



CaMKII-exo15 – SNP 2 (ATT deletion)
Heterozygous individuals: 010R

CaMK-new-exo - Vial 59 Inj 1 059 - Channel 1

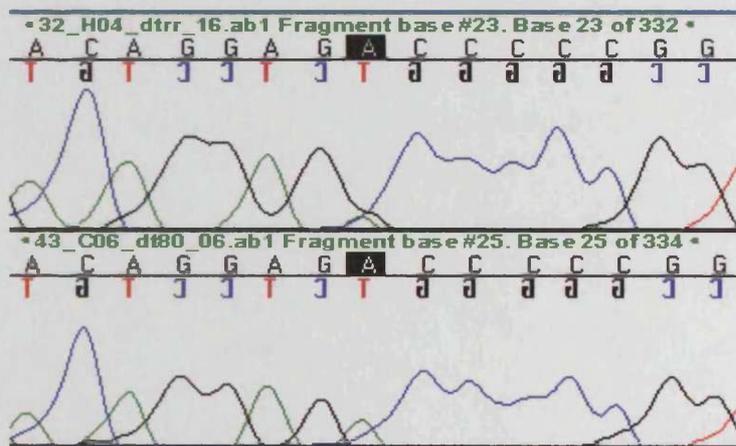
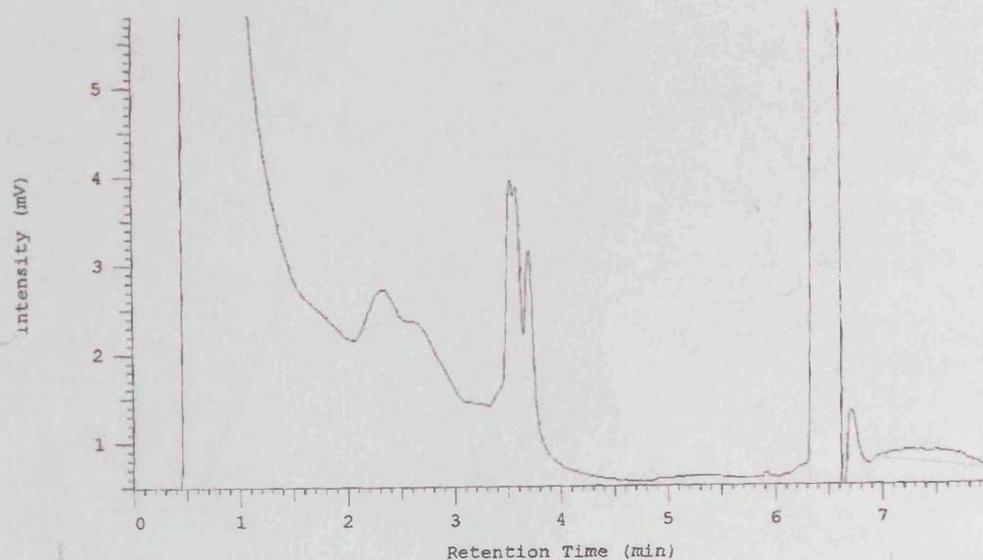
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Data Desc.: IFM CH1 2-D
Vial Number: 59 Inj Number: 1 Sample Name: 059



CaMKII-exo 17 – SNP 1 (G/A)
Heterozygous individuals: 001R

CaMK-new-exo - Vial 86 Inj 1 086 - Channel 1

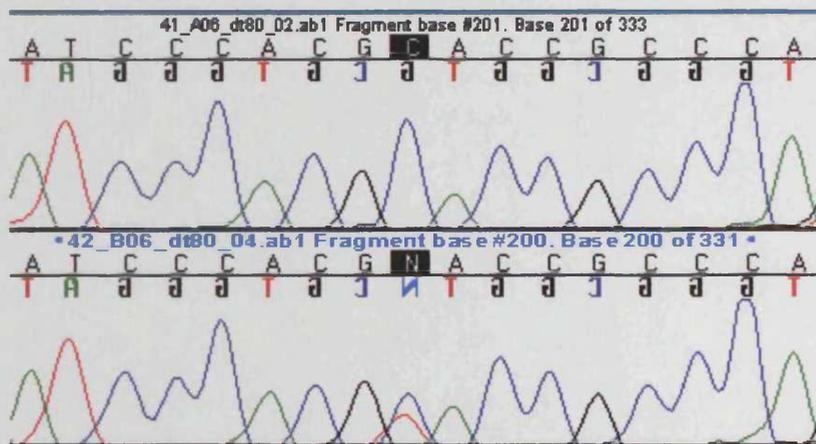
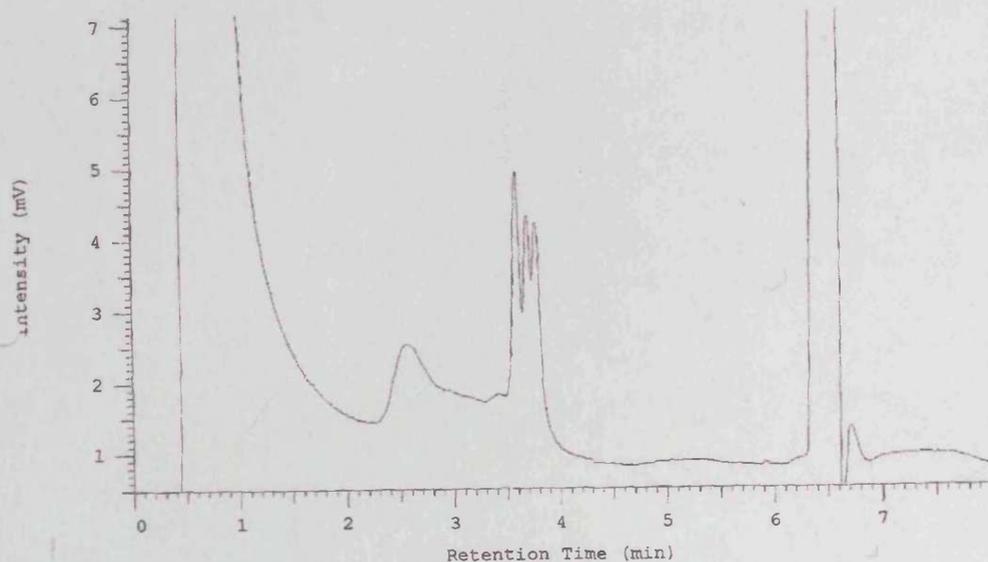
Current Data Path: D:\Dragana2\DATA\0153
Data Desc.: IFM CH1 2-D
Vial Number: 86 Inj Number: 1 Sample Name: 086



CaMKII-exo 17 - SNP 2 (C/T)
Heterozygous individuals: 015R

CaMK-new-exo - Vial 86 Inj 1 086 - Channel 1

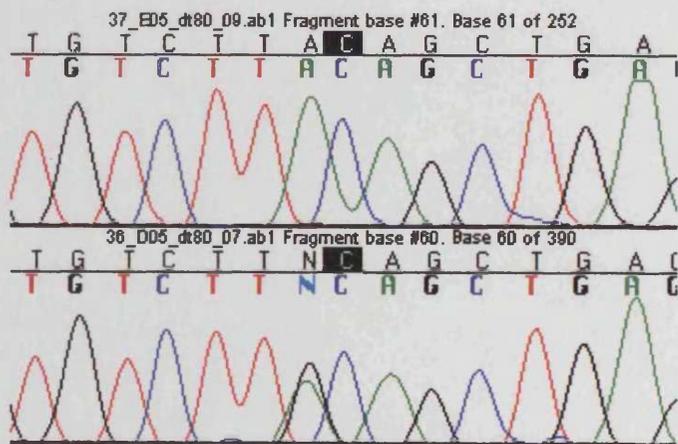
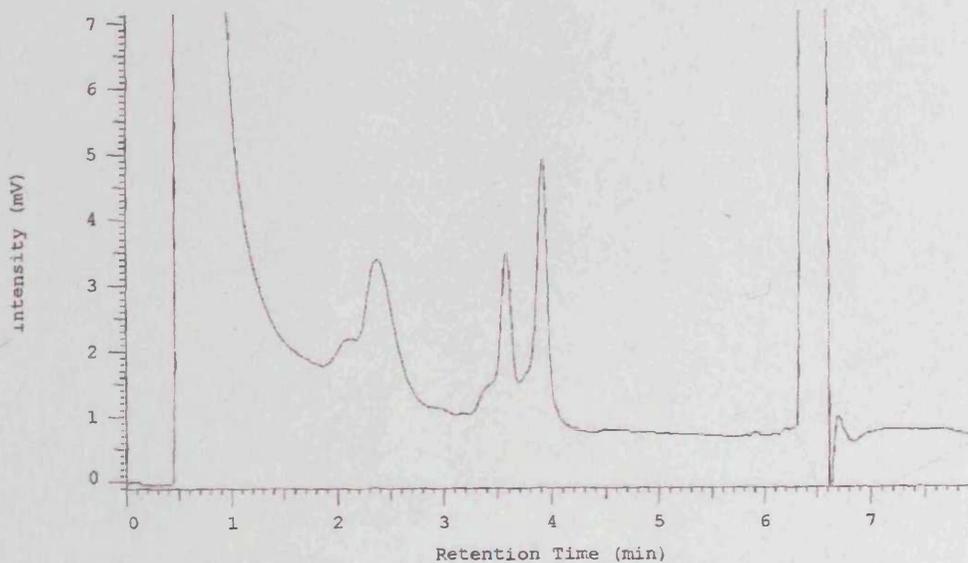
Current Data Path: D:\Dragana2\DATA\0148
Data Desc.: IFM CH1 2-D
Vial Number: 86 Inj Number: 1 Sample Name: 086



CaMKII-3'UTR-1- SNP 1 (A/G)
Heterozygous individuals: 004R

CaMK-new-3' - Vial 6 Inj 1 006 - Channel 1

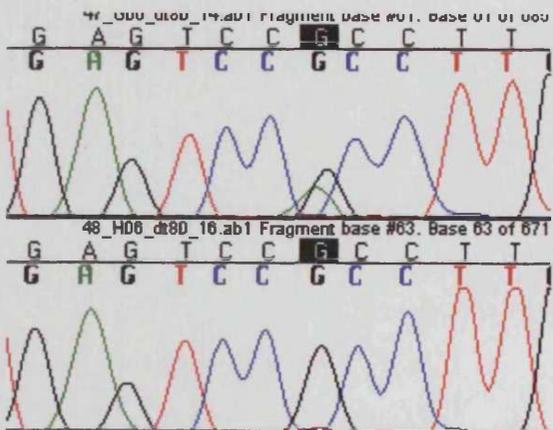
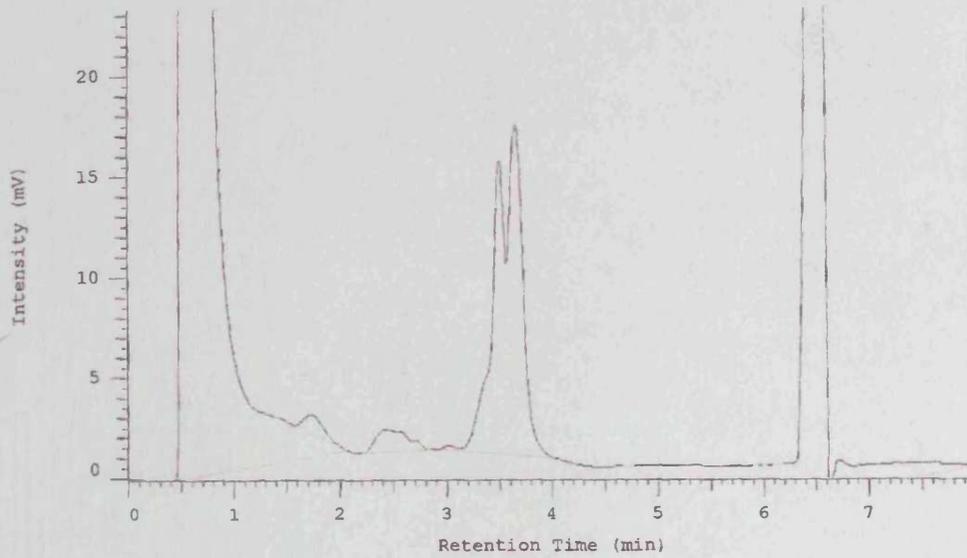
Current Data Path: D:\Dragana2\DATA\0160
Data Desc.: IFM CH1 2-D
Vial Number: 6 Inj Number: 1 Sample Name: 006



CaMKII-3'UTR-new 3 – SNP 1 (G/A)
Heterozygous individuals: 080R

CaMK-new-3'U - Vial 43 Inj 1 043 - Channel 1

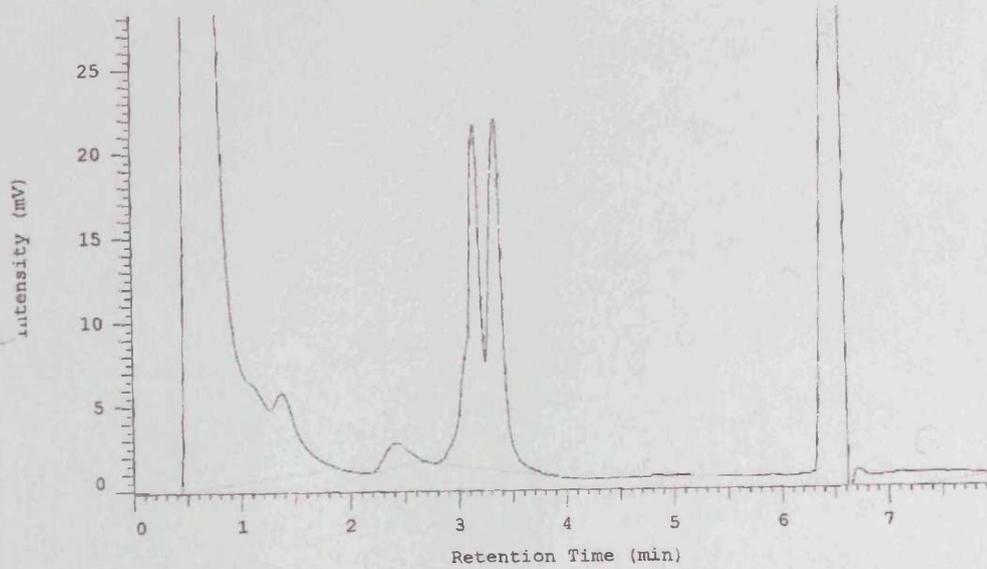
Current Data Path: D:\Dragana2\DATA\0187
Data Desc.: IFM CH1 2-D
Vial Number: 43 Inj Number: 1 Sample Name: 043



CaMKII-3'UTR-5- SNP 2 (A/G)
Heterozygous individuals: 009R

CaMK-new-3'U - Vial 75 Inj 1 075 - Channel 1

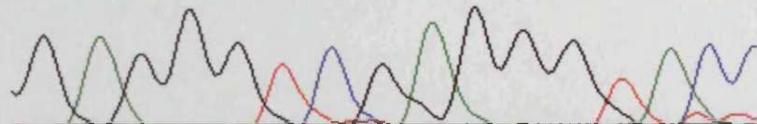
Current Data Path: D:\Dragana2\DATA\0190
Data Desc.: IFM CH1 2-D
Vial Number: 75 Inj Number: 1 Sample Name: 075



• 77_E10_dt180_10.ab1 Fragment base #195. Base 195 of 457 •
G A G G G T C N A G G G T A C C
G A G G G T C N A G G G T A C C



• 78_F10_dt180_12.ab1 Fragment base #195. Base 195 of 457 •
G A G G G T C G A G G G T A C C
G A G G G T C G A G G G T A C C



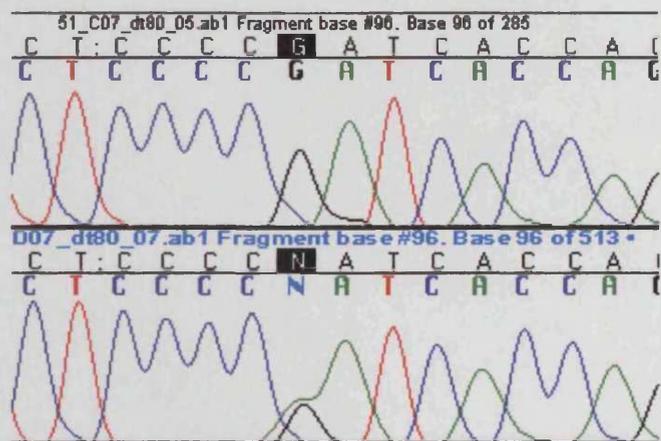
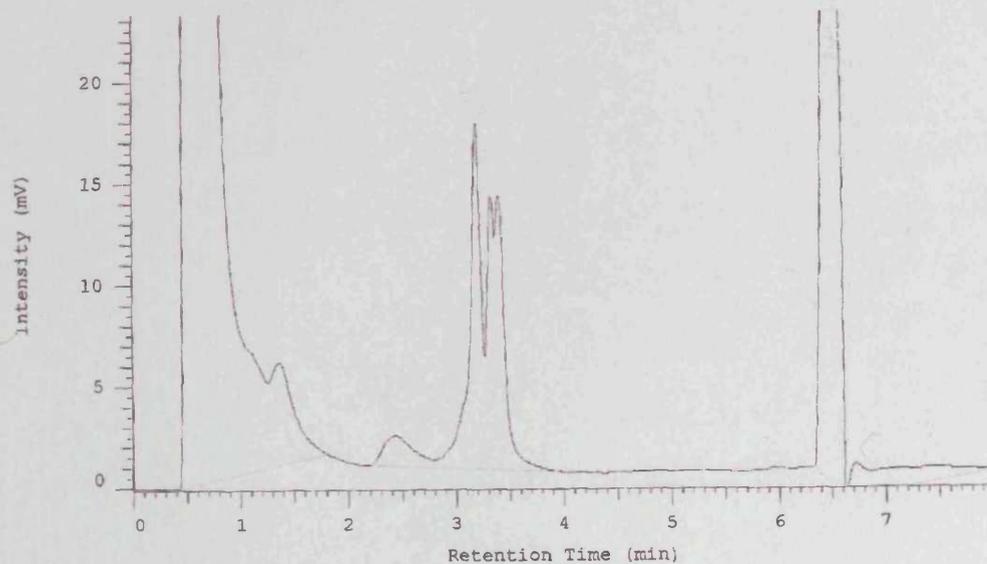
• 79_G10_dt180_14.ab1 Fragment base #195. Base 195 of 456 •
G A G G G T C A A G G G T A C C
G A G G G T C A A G G G T A C C



CaMKII-3'UTR-new 6 – SNP 1 (G/A)
Heterozygous individuals: 080R

CaMK-new-3'U - Vial 77 Inj 1 077 - Channel 1

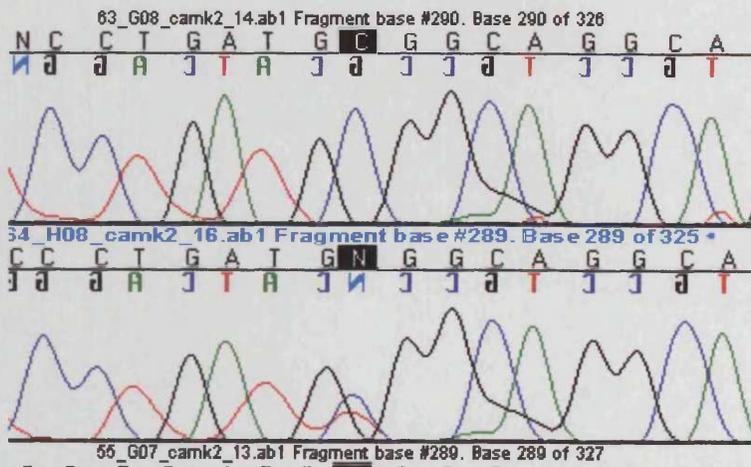
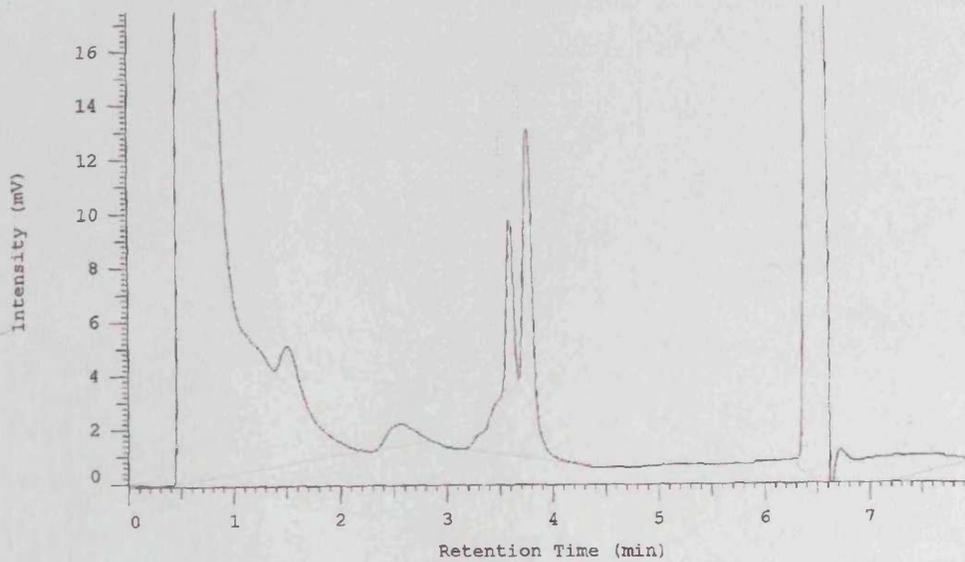
Current Data Path: D:\Dragana2\DATA\0190
Data Desc.: IFM CH1 2-D
Vial Number: 77 Inj Number: 1 Sample Name: 077



CaMKII-3'UTR- 7 – SNP 1 (C/T)
Heterozygous individuals: 001R

CaMK-new-3'U - Vial 91 Inj 1 091 - Channel 1

Current Data Path: D:\Dragana2\DATA\0191
Data Desc.: IFM CH1 2-D
Vial Number: 91 Inj Number: 1 Sample Name: 091



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