



UCLCHEM: A Gas-grain Chemical Code for Clouds, Cores, and C-Shocks*

J. Holdship¹, S. Viti¹, I. Jiménez-Serra², A. Makrymallis¹, and F. Priestley¹¹Department of Physics and Astronomy, University College London, Gower Street, WC1E 6BT, UK; jrh@star.ucl.ac.uk²School of Physics and Astronomy, Queen Mary University of London, 327 Mile End Road, London, E1 4NS, UK

Received 2016 October 31; revised 2017 May 16; accepted 2017 May 27; published 2017 July 5

Abstract

We present a publicly available, open source version of the time-dependent, gas-grain chemical code UCLCHEM. UCLCHEM propagates the abundances of chemical species through a large network of chemical reactions in a variety of physical conditions. The model is described in detail, along with its applications. As an example of possible uses, UCLCHEM is used to explore the effect of protostellar collapse on commonly observed molecules, and study the behavior of molecules in C-type shocks. We find the collapse of a simple Bonnor–Ebert sphere successfully reproduces most of the behavior of CO, CS, and NH₃ from cores observed by Tafalla et al. (2004), but cannot predict the behavior of N₂H⁺. In the C-shock application, we find that molecules can be categorized such that they become useful observational tracers of shocks and their physical properties. Although many molecules are enhanced in shocked gas, we identify two groups of molecules in particular. A small number of molecules are enhanced by the sputtering of the ices as the shock propagates, and then remain high in abundance throughout the shock. A second, larger set is also enhanced by sputtering, but then destroyed as the gas temperature rises. Through these applications, the general applicability of UCLCHEM is demonstrated.

Key words: astrochemistry – ISM: molecules – shock waves

1. Introduction

Chemistry is ubiquitous in astrophysical environments. Molecular clouds, the cold cores in which stars form, and the warm gas surrounding protostars all exhibit chemistry of varying degrees of complexity, with different dominant chemical pathways. Understanding this chemistry is vital to the study of our own origins, as well as understanding the physical structure and processes involved in star formation.

Beyond this, chemistry is a useful tool for understanding the physical conditions of the region being studied. This requires well-constrained chemical networks and accurate physical models, such that uncertainties in the predictions of the model are much smaller than the uncertainties in the measured abundances of molecules. With current state-of-the-art models, networks are generally capable of putting broad constraints on physical conditions such as maximum temperatures or minimum densities, which can be of use in poorly understood regions.

Chemical modeling is typically performed by the use of rate equations. The rates of a network of chemical reactions are calculated and used to determine the rate of change in a set of chemical species. This coupled set of ordinary differential equations (ODEs) is integrated to give the abundance of each species at any given time. These models typically center around gas-phase reactions provided by databases such as KIDA (Wakelam et al. 2012) and UMIST (McElroy et al. 2013). Each of these databases provides a chemical code, respectively Nahoon and RATE13, that solves these networks for simple gas conditions. They also include other processes such as H₂

formation on the dust grains and the self-shielding of CO and H₂ from UV radiation.

This, of course, does not account for all astrochemical processes, and many groups use more or less complicated chemical models for different purposes. In radiation-hydrodynamic models, it is common to reduce the network to gas-phased H, C and O based species, to reduce chemical integration time while reproducing the abundances of major coolants, such as CO, given by more detailed chemical models. On the other hand, the modeling of dense prestellar cores or the formation of complex organic molecules requires much larger chemical networks and the addition of processes involving dust grains.

For example, Astrochem (Maret & Bergin 2015) includes freeze-out of species onto dust grains and the non-thermal desorption of species from those grains due to UV radiation and cosmic rays, making it suitable for modeling a wider range of regions than a simple gas-phase model. By further including thermal desorption, star-forming regions with high gas/dust temperature conditions can be modeled. As the temperature of the core rises, the material on the grains sublimates and proper treatment of this sublimation is required. Astrochem and UCLCHEM implement these processes, with desorption from the grains occurring according to the binding energy of each species onto the grain.

However, Collings et al. (2004) demonstrated through temperature-programmed desorption experiments that multiple desorption events can occur for each species frozen on the dust grains, at different temperatures that are dependent both on the species and rate of temperature change. UCLCHEM additionally includes these desorption events for each species on the grain surface (See Section 2.1.4). The modeling of many protostellar environments benefits from the addition of grain chemistry, and proper temperature-dependent sublimation is required, for instance, for massive hot cores.

Another environment where proper grain surface and gas-phase chemical treatment have proven useful is in shocked gas. In these systems, sublimation from the grain surface is

* <https://uclchem.github.io>

dominated by sputtering, the collisional removal of surface material. The physical modeling of these shocks is an active area of research (e.g., Van Loo et al. 2009; Anderl et al. 2013) and codes for this modeling are publicly available (Flower & Pineau des Forêts 2015). These models have been successful in reproducing observations of shocked regions. For example, L1157-mm is a protostar driving a bipolar outflow (Gueth et al. 1997) and the bowshocks associated with that outflow are well-studied. For one of these bowshocks, L1157-B1, Gusdorf et al. (2008) reproduced the SiO emission using an MHD hydrodynamical model of a continuous or C-type shock, coupled with a reduced chemical network of 100 species.

As noted, reduced chemical networks often produce similar abundances to much more complex networks when simple species such as H₂O and CO are considered, and this is sufficient for many applications (e.g., Schmalzl et al. 2014). Similarly, a simplified physical model may reproduce shock features that are useful for chemistry, so a full chemical model can be run without fully solving the MHD equations. Jimenez-serra et al. (2008) produced a parameterized C-shock that produced a shock structure in good agreement with more complicated MHD models. This was combined with the chemical code UCLCHEM to study the chemistry in shocked environments (Viti et al. 2011). Using this parameterization with a large chemical network has yielded some success in the bowshock L1157-B1 (Viti et al. 2011; Holdship et al. 2016; Lefloch et al. 2016), despite being computationally inexpensive.

The simplicity of these models means they could be used to link observed molecular lines with the underlying chemistry to get a sense of the physical properties of the shock. Conversely, emission from a source with well-constrained physical properties could be used to improve uncertain parts of the chemical network. The reaction rates of species on dust grains being a prime example; e.g., Holdship et al. (2016) used observations of the L1157-B1 bowshock to constrain sulfur chemistry on ice grains by identifying likely sulfur carriers. Ideally, chemical models could be run over large parameter spaces to find the most likely values for uncertain parts of the network when a region has large amounts of observational data available.

To this purpose, UCLCHEM has been developed over many years, with numerous papers discussing major updates and applications (see Viti et al. 2004; Roberts et al. 2007). There have also been benchmarking efforts in the past (Viti et al. 2001), comparing the code to other similar models. We present here the most up-to-date³ version, the source code of which is now publicly available for the first time under the MIT license. This paper describes the code in its most recent state and provides example applications of the model. In Section 2 the code is discussed. Sections 3 and 4 provide two examples of the applications of UCLCHEM. The paper is summarized in Section 5.

2. UCLCHEM

UCLCHEM is a time-dependent gas-grain chemical model written in modern Fortran. It is primarily a chemical code, focusing on grain chemistry as well as gas-phase reactions. The chemistry includes freeze-out, thermal and non-thermal desorption, gas phase, and user-provided grain-surface-reaction network. In addition, hard-coded treatments for H₂ formation

on the grains and the self-shielding of CO and H₂ are contained in the chemistry module. The chemical solver calculates the rates of all the above processes to follow the fractional abundances of molecules for parcels of gas.

UCLCHEM makes use of modern Fortran’s modules to separate the chemistry and physics. Interchangeable physics modules control the number and physical conditions of gas parcels used in the chemistry to allow different scenarios to be modeled. In particular, the code is packaged with modules for molecular clouds, hot cores, C-shocks, and the post-processing of hydrodynamic simulations. In this section, the chemical model is explained, along with an overview of the physical models available and the Python code produced to create networks automatically.

2.1. Chemical Model

At its core, UCLCHEM is a code that sets up and solves the coupled system of ODEs that gives the fractional abundances of all the species in a parcel of gas. The ODEs are created automatically from the network (see Section 2.3). There is one ODE per species, each of which is a sum over the rates of every reaction that involves the species. At each time step, the rates of each reaction are recalculated and the third-party ODE solver DVODE (Brown et al. 1989) integrates to the end of the time step. Each type of reaction requires a different rate calculation, so UCLCHEM is limited by the kinds of reactions that have been coded. The subsections below detail each type of reaction and physical process currently included.

2.1.1. Gas-Phase Reactions

Gas-phase reactions make up the largest part of the chemical network. Two-body reactions, cosmic-ray interactions, and interaction with UV photons are all included in the code. The rate of each reaction is in cm³ s⁻¹. For two-body reactions, the rate is calculated through the Arrhenius equation,

$$R_{AB} = \alpha \left(\frac{T}{300 \text{ K}} \right)^\beta \exp\left(\frac{-\gamma}{T}\right), \quad (1)$$

where α , β , and γ are experimentally determined rate constants. For cosmic-ray protons, cosmic-ray induced photons, and UV photons the rates are given by

$$R_{\text{CRProton}} = \alpha_{\text{cr}} \zeta \quad (2)$$

$$R_{\text{CRPhoton}} = \alpha_{\text{cr}} \left(\frac{T}{300 \text{ K}} \right)^\beta \frac{E}{1 - \omega} \zeta \quad (3)$$

$$R_{\text{UV}} = \alpha_{\text{rad}} F_{\text{UV}} \exp(-kA_V), \quad (4)$$

where ω is the dust grain albedo, ζ is the cosmic ray ionization rate, F_{UV} is the UV flux, α_{rad} and α_{cr} scale the overall radiation field and cosmic ray ionization rates for a specific reaction, β takes the same meaning as above, E is the efficiency of cosmic ray ionization events, and k is a factor increasing extinction for UV light. This formulation of the reaction rates is taken from UMIST database notes; more information can be found in McElroy et al. (2013). These rates are then used to set up and solve a series of ordinary differential equations of the form,

$$\dot{Y}_{\text{product}} = R_{AB} Y_A Y_B n_{\text{H}}, \quad (5)$$

where Y is the abundance of reactants; A, B, and \dot{Y} are the rate of change of the product, and n_{H} is the hydrogen number

³ <https://uclchem.github.io>

density. The second abundance dependency is dropped for reactions between single species and photons or cosmic rays. An equivalent negative value is added to the change in the reactants.

In addition to these reactions, the rate of dissociation of CO and H₂ due to UV is reduced by a self-shielding treatment in the code. H₂ formation is treated with the hard-coded reaction rate,

$$R_{\text{H}_2} = 10^{-17} \sqrt{T} n_{\text{H}} Y_{\text{H}}, \quad (6)$$

where Y_{H} is the abundance of atomic hydrogen. This rate is taken from de Jong (1977) and performed well in chemical benchmarking tests (Röllig et al. 2007). It should be noted that, despite the self-shielding treatment and inclusion of UV photon reactions from UMIST, UCLCHEM is not a PDR code and should not be used to model regions where the UV field is sufficient to be considered a PDR. To model the chemistry in PDRs, the 3DPDR code described in Bisbas et al. (2012) is available on the UCLCHEM website, and a updated version of the 1D UCL_PDR (Bell et al. 2006) will be available soon.

2.1.2. Freeze Out

All species freeze-out at a rate given by

$$R_{\text{fr}} = \alpha_f \sqrt{\frac{T}{m}} a_g \text{fr}, \quad (7)$$

where a_g is the grain radius and fr is the proportion of each species that will freeze. Here, α_f is a branching ratio allowing the same species to freeze through different channels into several different species. Therefore, α_f allows the user to determine what proportion of a species will freeze into different products. This is used as a proxy for the relatively uncertain surface chemistry. For example, by freezing a portion of a species as a more hydrogenated species, rather than explicitly including a hydrogenation reaction in the surface network. The values for α_f in each freeze-out reaction should sum to 1 for any given species. Variable fr allows the user to set the freeze-out efficiency; it is left at 1.0 in this work, so that non-thermal desorption alone accounts for molecules remaining in the gas-phase. An additional factor C_{ion} is included for positive ions, to reproduce the electrostatic attraction to the grains,

$$C_{\text{ion}} = 1 + \frac{16.47 \times 10^{-4}}{a_g T}, \quad (8)$$

with a_g and T being the average grain radius and temperature, respectively, and the constant value taken from Rawlings et al. (1992).

2.1.3. Non-thermal Desorption

Three non-thermal desorption methods from Roberts et al. (2007) have been included in UCLCHEM. Molecules can leave the grain surface due to the energy released in H₂ formation, incident cosmic rays, and by UV photons from the interstellar radiation field, as well as those produced when cosmic rays ionize the gas.

$$R_{\text{DesH}_2} = R_{\text{H}_2\text{form}} \epsilon_{\text{H}_2} M_x, \quad (9)$$

$$R_{\text{DesCR}} = F_{\text{cr}} A \epsilon_{\text{cr}} M_x, \quad (10)$$

Table 1
List of Model Parameter Values

Variable	Value
Pre-shock Temperature	10 K
Initial Density	10^2 cm^{-3}
Freeze-out Efficiency	1.0
Radiation Field	1.0 Habing
Cosmic-ray Ionization Rate	$1.3 \times 10^{-17} \text{ s}^{-1}$
Visual Extinction at Cloud Edge	1.5 mag
Cloud Radius	0.05 pc
Grain Surface Area	$2.4 \times 10^{-22} \text{ cm}^2$
H ₂ Desorption (ϵ)	0.01
CR Desorption (ϕ)	10^5
UV Desorption (Y)	0.1

$$R_{\text{DesUV}} = F_{\text{uv}} A \epsilon_{\text{UV}} M_x, \quad (11)$$

where $R_{\text{H}_2\text{form}}$ is the rate of H₂ formation, F_{cr} is the flux of iron nuclei cosmic rays, and F_{uv} is the flux of ISRF and cosmic-ray-induced UV. The ϵ values are efficiencies for each process, reflecting the number of molecules released per event. A is the total grain surface area and M_x is the proportion of the mantle that is species x . The values of these parameters are given in Table 1, and the assumptions used to obtain them are discussed in Roberts et al. (2007).

2.1.4. Thermal Desorption

As the temperature of the dust increases, species start to desorb from the ice mantles. Laboratory experiments (e.g., Ayotte et al. 2001; Burke & Brown 2010) show that a species does not desorb in a single peak, but rather in multiple desorption events. These events correspond to the removal of the pure species from the mantle surface, a strong peak at the temperature corresponding to the binding energy, and the “volcanic” desorption and co-desorption of molecules that have diffused into the water ice. Collings et al. (2004) showed that many species of astrochemical relevance could be classified as CO-like, H₂O-like, or intermediate, depending on what proportion of the species is removed from the mantle in each desorption event.

To the authors’ knowledge, UCLCHEM is the only publicly available chemical model that implements a treatment for this. Molecules are classified according to their similarity to H₂O or CO, controlling their desorption behavior. Further, the user can now set the proportion of each molecule that enters the gas phase in each event, allowing the thermal desorption treatment of each molecule to improve with the laboratory data available. The user only needs to compile a list of species, along with the binding energies and desorption fractions for each grain species; the rest of this process is set up automatically (see Section 2.3). Although any physics module could activate the desorption process at specific temperatures, this process is strongly linked to the cloud model discussed in Section 2.2.1.

2.1.5. Grain Surface Reactions

Proper treatment of reactions between molecules on the grain surfaces is subject of debate. As a result, UCLCHEM has implemented grain chemistry in three ways. The method used in the basic network supplied with the code, and in the network used for the applications below, is to include hydrogenation at

freeze-out. For example, 10% of CO freezing onto the grains is assumed to go on to form CH₃OH in the network used here, so it is immediately added to the grain abundance of CH₃OH.

More complicated treatments are possible without editing the code. By including reactions with modified version of the rate constants in the network, grain surface reactions can be treated in the same way as gas-phase reactions. Further, UCLCHEM calculates the rates of reactions by diffusion across the grain surfaces for reactions with a third reactant labelled “DIFF.” In this case, the binding energy of the two actual reactants are used to calculate their diffusion rates and included in a modified rate equation according to the treatment of Occhiogrosso et al. (2014).

2.2. Physical Model

A number of physics modules are included with the code and can be interchanged as required. They are explained briefly below.

2.2.1. Clouds and Collapse Modes

The “cloud” module is the main physics module of UCLCHEM. This sets up a line of parcels from the center to the edge of a cloud of gas and controls the density, temperature and visual extinction of each effectively giving a 1D model of a cloud. The conditions at the outer edge of the cloud are set by an interstellar radiation field and a value of visual extinction at the cloud edge. The distance from the edge of the cloud and the densities of the parcels closer to the edge are used to calculate the visual extinction and hence the radiation field for each interior parcel but parcels are otherwise treated separately.

The model works in two phases. In phase 1, this module is most often used to follow the collapse of the cloud from a diffuse medium to the density required for phase 2. Starting from elemental abundances only, this produces a set of gas and grain molecular abundances that is self-consistent with the network. These can be used as the starting values for phase 2, rather than assuming a set of initial abundances.

In phase 2, the temperature increases as the cloud collapses, so the envelope around a forming protostar can be modeled. As discussed above, sublimation of species from the grain is dependent both on the temperature and its rate of change. In the cloud model, the temperature of the gas as a function of time and radial distance from the protostar is calculated at each time step, then compared to pre-calculated temperatures to determine when major thermal desorption events should occur. The temperature profiles and desorption peaks are from Viti et al. (2004), where the time-dependent temperature profiles and desorption temperatures for a range of high stellar masses were calculated. This has since been extended with radial dependency and lower masses. Alternatively, a module that reads and interpolates the output of hydrodynamical codes can be used to post-process the chemistry of hydrodynamical models.

Phase 1 can also be used to study cold gas. Gas in a steady state can be modeled by setting all the required parameters and turning off collapse. Further, a number of different collapse modes have been coded into the model. The standard collapse used to create a cloud for phase 2 is the freefall collapse taken from Rawlings et al. (1992). Parameterizations for the collapse of a Bonnor–Ebert sphere are also included, so that the collapse of cold cores can be studied. These are described in more detail by F. Priestley et al. (2017, in preparation). Each collapse model calculates the density of a parcel by following the central

density of the Bonnor–Ebert sphere with time, then scaling for the radial distance of a parcel from the center of the core. The chemistry of each parcel is evolved separately, such that chemical abundances as a function of radius can be studied for different collapse modes and compared to observations of collapsing cores.

2.2.2. C-shock

The “C-shock” model is the parameterized model of continuous or C-type shocks from Jimenez-serra et al. (2008), adapted for use in UCLCHEM. The parameterization calculates the evolution of the velocities of the ion and neutral fluids as a function of time, and deduces the physical structure of the shock from simple principles and approximations. In addition to the changes in density and temperature, the shock model includes a treatment of sputtering. The saturation time, t_{sat} , is defined in Jimenez-serra et al. (2008). It corresponds to the time at which the abundances of species sputtered from the icy mantles change by less than 10% for two consecutive time steps in their calculations. This gives a measurement of the timescales at which most of the material that should be sputtered from dust grains will have been released. As the timescales in which the mantles sputter are short under the typical conditions of molecular clouds (Jimenez-serra et al. 2008), this is treated by releasing all of the mantle material into the gas phase at t_{sat} . The equations used to calculate the saturation time do not depend on molecular mass, so t_{sat} is the same for all species. See Jimenez-serra et al. (2008) Appendix B for more details.

Although sputtering is included, there are two major grain processes that are omitted. Grain-grain collisions in the shock can cause both shattering and vaporization. Shattering is the breaking of grains in collisions; it primarily has the effect of changing the dust-grain size distribution. This produces hotter and thinner shocks, particularly for shocks with higher speeds or weak magnetic fields (Anderl et al. 2013). Vaporization is the desorption of mantle species due to grain heating from collisions. The inclusion of grain–grain processes vastly increases the complexity of the code, and would not allow the code to be run so quickly with low computing power. Shattering is particularly likely to greatly affect the abundance of Si-bearing species, as the destroyed dust grains would add to the gas phase abundances. It should be noted however, that in models where SiO is included in the mantle, Anderl et al. (2013) find SiO abundances that are consistent with this model. The interested reader is referred to recent work on grain–grain processes in shocks for a more detailed view of this (Guillet et al. 2010; Anderl et al. 2013).

The model discussed here has been successfully used to explain the behavior of molecules like H₂O, NH₃, CS, and H₂S in shocks like L1157-B1 (Viti et al. 2011; Gómez-Ruiz et al. 2014; Holdship et al. 2016). The parameterization has been extensively tested against MHD C-shock simulations, and performs well across a wide range of input parameters. Importantly, it is computationally simple. This allows the chemistry to be the main focus without computational costs prohibiting us from running large grids of models to compare quickly to observations.

2.3. Network-makerates

A network of all the included species and their reactions must be supplied to the model. For UCLCHEM, this consists of two parts: the gas-phase reactions taken from the UMIST

database and a user-defined set of grain reactions. Grain reactions include freeze-out, non-thermal desorption, and any reactions between grain species. The UCLCHEM repository includes the UMIST12 data file (McElroy et al. 2013) and a simple set of grain reactions. Given that the code must be explicitly given the ODEs and told which species to sublimate, and in which proportions for different thermal desorption events, a Python script has been created to produce the input required to run UCLCHEM from a list of species, a UMIST file, and a user-created grain file.

3. Application I: Collapsing Cores

To provide an example of the capabilities of UCLCHEM, the results of a model run for a collapsing core of gas is presented here. In the earliest stages of star formation, protostars form from collapsing cores of gravitationally bound gas. However, the evolution of the density profiles of these cores is not well-understood. UCLCHEM can be used with a variety of theoretical time-dependent density profiles. In a separate paper, F. Priestley et al. (2017, in preparation) will discuss these profiles and their chemical effects in detail. Here, a simple and commonly considered profile is demonstrated, that of a marginally super-critical Bonnor–Ebert (BE) sphere (Bonnor 1956). The link to the code provided in previous sections will contain the most up-to-date release of UCLCHEM; a permanent record of the code used for this work is available at (Holdship 2017, Codebase: <https://doi.org/10.5281/zenodo.580044>).

BE spheres are a solution to the hydrostatic equilibrium equations; they have been shown to fit the observed density profiles of prestellar cores. This typically requires that the cores are close to the critical mass that would make them unstable to collapse (Kandori et al. 2005). The cloud model includes a parameterization of simulations by Foster & Chevalier (1993), who consider a BE sphere at the critical mass for stability, then increase the density throughout the sphere by 10%, making it marginally super-critical. Thus, the sphere collapses and the chemistry of a prestellar core collapsing in this fashion can be modeled. Such applications are useful as it may be possible to discriminate between different collapse models from the observed chemical differentiation in the core, even if the density profile cannot be confidently described due to the lack of a sufficiently high angular resolution.

3.1. Model Setup

The model runs consider 50 gas parcels equally distributed from the core to the edge of a cloud of size 0.2 pc, with an initial density of $2 \times 10^2 \text{ cm}^{-3}$. This cloud is embedded in a larger medium, which adds 1.5 mag to the visual extinction at the cloud edge. This cloud maintains a temperature of 10 K as it collapses, according to the freefall equation

$$\frac{dn}{dt} = \left(\frac{n_{\text{H}}^4}{n_0}\right)^{\frac{1}{3}} \left[24\pi G m_{\text{H}} n_0 \left(\left(\frac{n_{\text{H}}}{n_0}\right)^{\frac{1}{3}} - 1 \right) \right]^{\frac{1}{2}}, \quad (12)$$

where n_{H} is the hydrogen nuclei number density, n_0 is the initial central density, and m_{H} is the mass of a hydrogen nucleus. The initial molecular abundances are set to zero, and elements are set to solar abundance values (Asplund et al. 2009). The cloud then collapses until it reaches a density of $2 \times 10^4 \text{ cm}^{-3}$, the initial

density used for the BE collapse. This takes approximately 3.5 Myr. The chemical processes described in Section 2.1 occur during this collapse. This process is typically referred to as phase 1 in work using UCLCHEM, and is used to create initial abundances and physical conditions for the following phase, which is typically the process of interest. In this case, the abundances at the end of this collapse are used as the initial abundances for model runs that start at $2 \times 10^4 \text{ cm}^{-3}$ and collapse to $1 \times 10^8 \text{ cm}^{-3}$, either in freefall collapse or according to the BE collapse parameterization.

3.2. Results

In order to test the model, it is useful to compare to observations. Tafalla et al. (2004) (T04, hereafter) present observations of two starless cores, L1517B and L1498, that are well-approximated by BE spheres with central densities of approximately $2.2 \times 10^5 \text{ cm}^{-3}$ and $0.94 \times 10^5 \text{ cm}^{-3}$, respectively. Although the reported characteristics for each core are very similar, L1498 is here used as a comparison to the model. T04 derive temperature and density profiles from the dust emission. They then derive CS, CO, NH_3 , and NH_2^+ abundance profiles from radiative transfer fitting to the flux profile, favoring the simplest profile that recreated the data.

They find CO and CS to be depleted in the denser centers of the cores compared to the outer radii, and that this is adequately represented as a step function. In contrast, NH_3 is enhanced by a factor of a few in the center; a trend they represent as a simple radial power law. N_2H^+ has a constant abundance for L1517B, but a slight drop at large radii for L1498. Chemical models including the one presented here accurately reflect the behavior of CS and CO. However, many struggle to reproduce the NH_3 and N_2H^+ behavior (Aikawa et al. 2003). Figure 1 shows results of UCLCHEM with the BE model used to fit these observations at a central density of $0.95 \times 10^5 \text{ cm}^{-3}$.

From Figure 1, it can be seen that the BE sphere modeling with UCLCHEM yields some success. For CO and CS, T04 find each can be treated as having constant abundance, outside of some cutoff radius, with the abundance in the core being greatly depleted. For simplicity, T04 take a central abundance of zero; although this is clearly not the case for UCLCHEM, there are some caveats. The T04 observations are of C^{17}O and C^{18}O , and they report C^{18}O abundances. For the comparison here, those values have been multiplied by 550, the $\text{C}^{16}\text{O}:\text{C}^{18}\text{O}$ ratio (Wilson 1999). Reducing the CO values by this amount gives a central abundance of 2×10^{-10} , which is sufficiently low that the signal should be too weak to detect. Similarly, the central abundance of CS in the UCLCHEM model is 4×10^{-11} , and therefore in agreement with zero.

T04 provide an analytic expression for NH_3 which is plotted in Figure 1; the main finding is that NH_3 increases by a factor of a few toward the center of the core. The UCLCHEM model predicts the opposite, increasing with radius. UCLCHEM similarly struggles to reproduce observations of N_2H^+ , decreasing both toward the center and edge of the core, rather than the approximately constant abundance found in T04. It appears from the model outputs that, at middle radii, N_2H^+ is forming through a reaction between H_3^+ and N_2 . The increased density of the middle radii makes this reaction more efficient than it is at the edges, but the even-higher density and

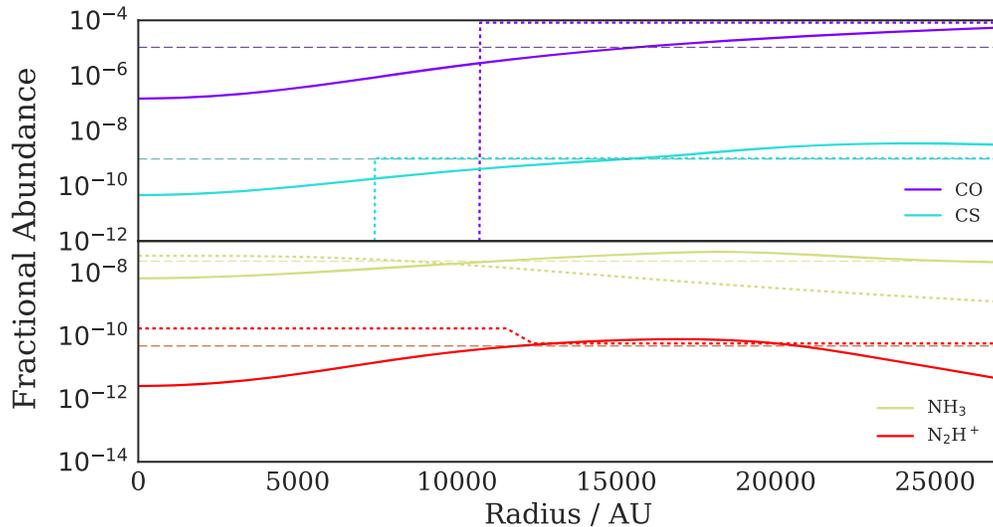


Figure 1. Fractional abundance as a function of radius for molecules from Tafalla et al. (2004) modeled by UCLCHEM are plotted as solid lines. Dashed, horizontal lines plot the results of a freefall model, to demonstrate the improvements to the model achieved by using a BE sphere collapse.

visual extinction of the center reduces the amount of H_3^+ available, and further increases the rate of freeze-out.

The freefall model is plotted on the same figure, using thinner dashed lines. The freefall model only takes into account the initial density of each parcel, not the radius, so there is no differentiation with distance from the core center. Further, the freefall model reaches $1 \times 10^5 \text{ cm}^{-3}$ much more quickly than the BE model, so the depletion of each species is much lower than expected because freeze-out has less time to occur. In many situations, the freefall model is sufficient for single-point models that describe the densest part of a cloud of gas, but it is clearly inappropriate to model the collapse of prestellar cores.

The shortcomings of the BE model could be due to a number of factors. Spherical symmetry is assumed in both the UCLCHEM model and the radiative transfer modeling of T04, who fit to azimuthally averaged data. Alternatively, the collapse of a marginally super-critical BE sphere may not be exactly appropriate for this object, or the initial conditions for the chemical model may be incorrect. The fact the behavior of both N_2H^+ and NH_3 are not reproduced by the BE model could indicate a problem with the nitrogen chemistry in the dense gas at core center. However, the model is clearly superior to a simple freefall collapse, and a more thorough investigation into collapse modes and initial conditions could link the observed abundance profiles and the chemical modeling. The strength of UCLCHEM is that it has been specifically designed to have physics models that are easily modified, and thus easily implemented. To illustrate the code’s range of physical applications, in the next section, we present a second example with the chemical modeling of C-type shocks.

4. Application II: Shock Chemistry

4.1. How Can We Trace Shocks?

In this section, we present a second example use of UCLCHEM that focuses on the study of C-type shocks. A primary concern of chemical modeling is the identification of tracer molecules, the observation of which can provide information on a quantity that is not directly observed. We aim to find molecules that trace shocks, particularly those sensitive to the shock properties.

Here, we build on the work of Viti et al. (2011), in which H_2O and NH_3 lines were used to determine the properties of a shock. Some molecules, such as H_2O , trace the full shock because they are not destroyed in the hotter parts of the post shock gas, which then dominates the high-J emission. Others, such as NH_3 , may not; they are instead destroyed in the hotter parts of the post-shock gas. This behavior depends on both the pre-shock density and the shock speed, so the determination of which molecules fall into which category in varying conditions is a rich source of information. We provide groupings of molecules under different shock conditions (pre-shock density and shock speed) so comparisons can be made between groups when observing shocked gas.

4.2. Model Setup

As explained in previous sections, the model is two-phased. Phase 1 starts from purely atomic gas at low density ($n_{\text{H}} = 10^2 \text{ cm}^{-3}$ in this work), which collapses according to the freefall equation, in a manner similar to the core collapse model described above. This sets the starting abundances for all molecules in phase 2, which will be self-consistent with the network, rather than being assumed values for a molecular cloud. The elemental abundances are inputs; for this work, typical elemental solar abundances from Asplund et al. (2009) were used. The cloud collapses to the initial density of phase 2 and is held there until a chosen time. The grid run for this work used a phase 1 of 6 Myr, which provided sufficient time for the required pre-shock densities to be reached.

Table 1 lists the physical parameters that were set for all models run for this work. The phase 1 models used a temperature of 10 K, a radiation field of 1 Habing, and the standard cosmic-ray ionization rate of $1.3 \times 10^{-17} \text{ s}^{-1}$. A radius of 0.05 pc is chosen for the cloud; this sets the A_{v} by allowing the column density between the edge of the cloud and the gas parcel to be calculated assuming a constant density. An additional 1.5 mag are added to the A_{v} , an assumed value for the amount of extinction from the interstellar radiation field to the edge of the cloud. The final density of this phase varies, depending on the density required for phase 2 (See Table 2).

Table 2
List of Shock Variable Values

Model	$\log(n_{\text{H},0})$	V_s /km s ⁻¹	t_{sat} /yr	T_{max} /K	Z_{diss}^a /cm
1	10 ³	10	975.7	323	3.70×10^{17}
2	10 ⁴	10	97.6	323	3.70×10^{16}
3	10 ⁵	10	9.8	301	3.70×10^{15}
4	10 ⁶	10	1.0	279	3.70×10^{14}
5	10 ³	15	746.5	584	5.55×10^{17}
6	10 ⁴	15	74.7	584	5.55×10^{16}
7	10 ⁵	15	7.5	555	5.55×10^{15}
8	10 ⁶	15	0.8	525	5.55×10^{14}
9	10 ³	20	586.13	869	7.41×10^{17}
10	10 ⁴	20	58.61	869	7.41×10^{16}
11	10 ⁵	20	5.86	893	7.41×10^{15}
12	10 ⁶	20	0.59	916	7.41×10^{14}
13	10 ³	25	483.09	1178	9.26×10^{17}
14	10 ⁴	25	48.31	1178	9.26×10^{16}
15	10 ⁵	25	4.83	1316	9.26×10^{15}
16	10 ⁶	25	0.48	1454	9.26×10^{14}
17	10 ³	30	425.83	1510	1.11×10^{18}
18	10 ⁴	30	42.58	1510	1.11×10^{17}
19	10 ⁵	30	4.26	1824	1.11×10^{16}
20	10 ⁶	30	0.43	2137	1.11×10^{15}
21	10 ³	35	402.8	1866	1.30×10^{18}
22	10 ⁴	35	40.28	1866	1.30×10^{17}
23	10 ⁵	35	4.03	2416	1.30×10^{16}
24	10 ⁶	35	0.4	2966	1.30×10^{15}
25	10 ³	40	402.47	2245	1.48×10^{18}
26	10 ⁴	40	40.25	2245	1.48×10^{17}
27	10 ⁵	40	4.02	3093	1.48×10^{16}
28	10 ⁶	40	0.4	3941	1.48×10^{15}
29	10 ³	45	413.29	2648	1.67×10^{18}
30	10 ⁴	45	41.33	2648	1.67×10^{17}
31	10 ⁵	45	4.13	3855	1.67×10^{16}
32	10 ³	60	397.28	3999	2.22×10^{18}
33	10 ⁴	60	39.73	3999	2.22×10^{17}
34	10 ³	65	337.32	4497	2.41×10^{18}
35	10 ⁴	65	33.73	4497	2.41×10^{17}

Note.

^a Z_{diss} , the dissipation length, is the length scale over which the shock dissipates, beyond which the model is no longer applicable.

In phase 2, the C-shock occurs. Table 2 gives a full list of the C-shock models that were run. Not all of the listed variables are independent; t_{sat} and T_{max} are functions of the pre-shock density ($n_{\text{H},0}$) and shock velocity (V_s). The T_{max} , V_s relationship is extracted from Figures 8 and 9 of Draine et al. (1983), which give values of maximum temperature for pre-shock densities 10⁴ and 10⁶ cm⁻³. Because no values are available when the pre-shock density is 10⁵ cm⁻³, the maximum temperature is calculated as the average of the 10⁴ and 10⁶ values. More details of these variables and how they are calculated can be found in Jimenez-serra et al. (2008). The temperature and density profiles of the gas during this C-shock are shown in Figure 2, using model 27 as an example.

4.3. Shock Tracers

Due to the fact freezing is efficient in cold, dense clouds, many species are highly abundant in the solid phase of the pre-shock cloud model used in this study. The result is that the fractional abundances of a similarly large number of species increase by orders of magnitude in shocks of any speed, and

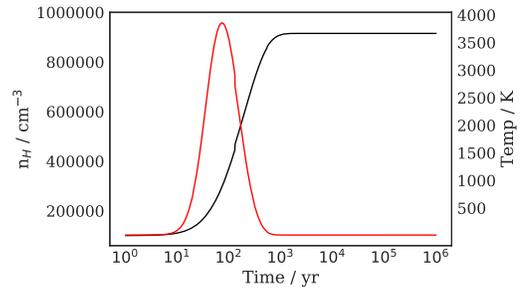


Figure 2. Neutral fluid temperature and density profiles of a typical model. In this case, model 27 is shown. Temperature is shown in red and the density in black.

can be used to determine whether a shock has passed. These are typically molecules that do not have efficient formation mechanisms in cold gas, but are easily formed on the grain. In particular, hydrogenated molecules tend to fall into this category.

H₂O and NH₃ are highly abundant in the ices, but not the gas phase of cold molecular clouds. Of the two, NH₃ is a more useful tracer of shocks due to the difficulties in observing H₂O from the ground. In the models, NH₃ has a fractional abundance of 10⁻¹⁴ in the pre-shock gas, but can increase to 10⁻⁵ in a shock (see Figure 3). H₂S and CH₃OH similarly go from extremely low abundances in the pre-shock gas to $X \sim 10^{-5}$ in a shock. Fractional abundances as a function of shock speed for different pre-shock densities are shown in Figure 3.

An important caveat to this is that the model assumes all of the grain surface material is released unchanged into the gas phase. This may not be the case, because the collisions that cause sputtering may have enough energy to destroy molecules as they are released. Suutarinen et al. (2014) discuss this possibility in the context of CH₃OH being released from ice mantles. They find that more than 90% of CH₃OH is destroyed either in the sputtering process itself or in gas-phase reactions after sublimation. The latter is the focus of the next section of this work, but it is not clear how much each process contributes to the destruction. However, the species presented here increase by over six orders of magnitude in gas-phase abundance, due to sputtering, if even 1% survives the process then the enhancement is observable.

An attempt was made to identify species that vary in fractional abundance by orders of magnitude depending on the shock speed. This is in contrast to those shown in Figure 3, which show a steep change from no shock to a low-velocity shock, but no variation thereafter. Such species would allow for the possibility of determining shock properties directly from the abundance of certain molecules. Despite many species increasing in shocks, it was not possible to identify any such species. In the model, even the low-velocity shocks (10 km s⁻¹) sputter the grains such that all shocks can increase gas-phase abundances in this way. Any variability due to gas temperatures reached in the shock is lost when an average over the whole shock is taken, with abundances varying by less than an order of magnitude.

It may not be possible to use the abundances directly to determine the shock velocity. However, UCLCHEM provides a useful input for the radiative transfer: a gas-phase abundance for the molecule as a function of the passage of the shock. Further, as discussed in Section 4.4, although bulk abundances

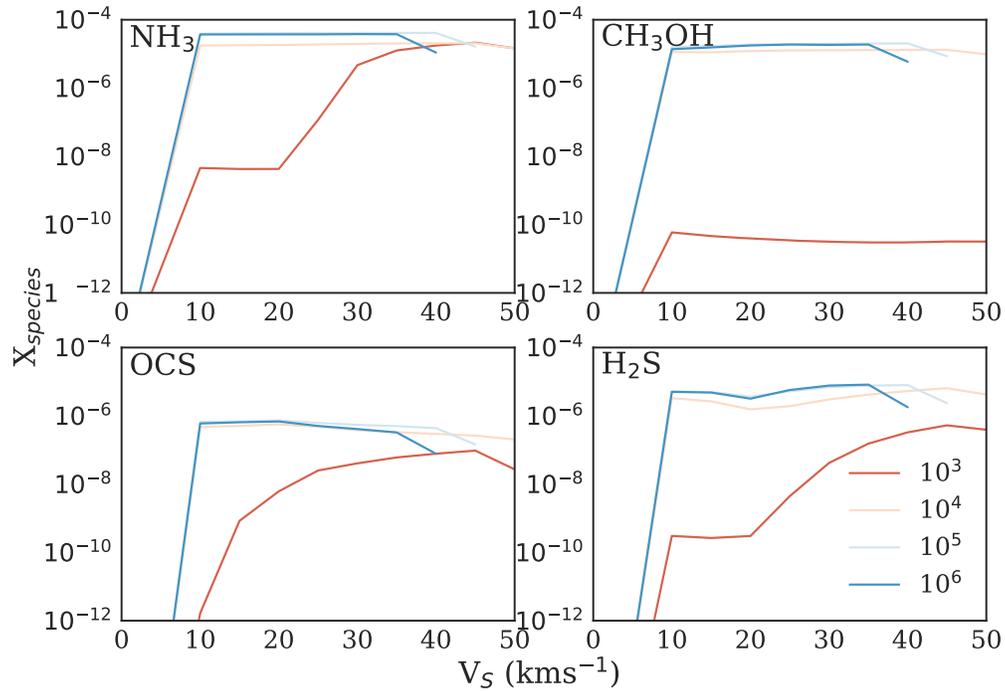


Figure 3. Average fractional abundance as a function of shock velocity for a number of molecules that have low gas-phase abundances in the pre-shock gas, but show high abundance increases in shocks of any velocity. Each line shows a different pre-shock density; the $\log(n_H)$ values are given in the lower right plot. In the absence of other mechanisms capable of sublimating most of the species in the ices, high abundances of these molecules would indicate the passage of a shock.

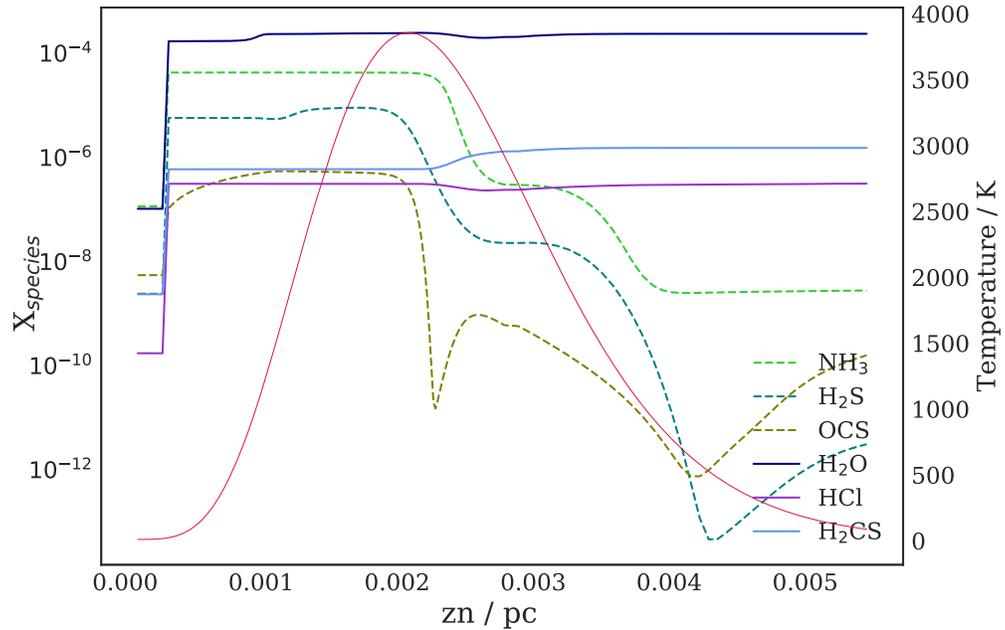


Figure 4. Fractional abundances as a function of distance into a shock for model 31. Species from each of the two groups are displayed as an illustration of their differences. The solid lines show H₂O-like molecules and the dashed ones show NH₃-like molecules. The neutral gas temperature profile is marked in red to show where the NH₃-like molecules start to fall in abundance.

are not sensitive to shock speed, line profiles can be used to provide further constraints.

4.4. Determining Shock Properties

The focus so far has been on average abundances across the whole shock, the observed abundance if one only resolves the shocked region. However, molecules can exhibit differences through the shock at different velocities, even when the average

is unchanged. If these differences are ignored, the models can only be used to infer that a shock has passed. However, if the velocity profile of the emission is taken into account, much more can be learned from the models.

For example, Codella et al. (2010) presented NH₃ and H₂O profiles from the L1157-B1 region. They showed that, when scaled to be equal at their peak emission velocity, NH₃ dropped off much more quickly with increasing velocity. Viti et al. (2011) demonstrated that this is consistent with NH₃ being at

Table 3
Shock Tracers for All Initial Density (Vertical Values) and Shock Speed (Horizontal Values) Pairs from Table 2

	10	15	20	25	30	35	40	45
	320	580	870	1200	1500	1900	2200	2600
10^3	H₂O HCl	H₂O HCl	H₂O HCl	CS	CS	C₃H₂ H ₂ O HCl HS H ₂ CN CS	C₃H₂ H ₂ O HCl HCN HS H ₂ CN	C₃H₂ H ₂ O HCl HCN HS H ₂ CN
	320	580	870	1200	1500	1900	2200	2600
10^4	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS H ₂ CO H ₂ S O ₂ HCO HS NS	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS H ₂ CO H ₂ S O ₂ HCO HS NS	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS NS H ₂ S O ₂ CN HNO OH HCO HS NS	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS NS H ₂ S O ₂ CH CN HNO OH HCO HS NS	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ NS H ₂ S O ₂ CH CN HNO OH HCO HS NS H ₂ CO	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ NS H ₂ S O ₂ CH CN HNO OH HCO HS NS H ₂ CO	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ NS H ₂ S HS CH CN HNO HCO NS H ₂ CO O₂	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ NS H ₂ S HS CH CN HNO HCO NS H ₂ CO O₂
	300	560	890	1300	1800	2400	3100	3900
10^5	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS O ₂ SO HNC HCN HCO	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS O ₂ SO HNC HCN HCO	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS O ₂ SO HNC HCN HCO	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS O ₂ SO HCN	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS O ₂ SO HCN HNC	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS SO HCN O₂ HNC	H ₂ O HCL H ₂ CS CH ₃ OH NH ₃ OCS SO HCN O₂ HNC H ₂ CO	H ₂ O HCL H ₂ CS HCN O₂ HNC H ₂ CO CO ₂ H ₂ S NH₃ HNO HCO SO OCS CH₃OH
	280	530	920	1500	2100	3000	3900	
10^6	H ₂ O HCL H ₂ CS CH ₃ OH H ₂ S NH ₃ OCS CO ₂ O ₂ SO HNC HCN HCO	H ₂ O HCL H ₂ CS CH ₃ OH H ₂ S NH ₃ OCS CO ₂ O ₂ SO HNC HCN HCO	H ₂ O HCL H ₂ CS CH ₃ OH H ₂ S NH ₃ OCS CO ₂ O ₂ SO HCN HCO OH	H ₂ O HCL H ₂ CS CH ₃ OH H ₂ S NH ₃ OCS CO ₂ O ₂ SO HCN HCO OH	H ₂ O HCL H ₂ CS CH ₃ OH H ₂ S NH ₃ OCS CO ₂ SO HCN HCO OH	H ₂ O HCL H ₂ CS CH ₃ OH H ₂ S NH ₃ OCS CO ₂ SO HCN O₂ HNC OH H ₂ CO	H ₂ O HCL H ₂ CS HCN NH₃ O₂ CH ₃ OH HNC H ₂ CO CO ₂ H ₂ S HNO HCO SO OCS	

Note. The maximum gas temperature reached in each model is displayed in the relevant box. Species in blue are enhanced by the shock and trace its full extent. Species in red are enhanced initially by the shock, but destroyed as the shock increases the temperature. Species in bold are those that exhibit a given behavior only for a small range of conditions, making them particularly useful for determining shock properties. The bottom right panel is left blank; a C-type shock cannot propagate at 45 km s⁻¹ through a medium of density 10⁶ cm⁻³.

lower abundance in the hotter gas, with chemical models showing that a shock with $v_s > 40$ km s⁻¹ and pre-shock density of $n_H = 10^5$ cm⁻³ was sufficient to destroy NH₃ (which, unlike H₂O, does not efficiently reform in the gas phase). As such, a divergence in the velocity profiles of these molecules can indicate such a shock.

Gómez-Ruiz et al. (2016) took this further and showed that the difference in the terminal velocity of NH₃ and H₂O was correlated with the shock velocity inferred from the H₂O terminal velocity for a sample of protostellar outflows. Chemical modeling in the work corroborated this by showing more NH₃ being lost in faster shocks. Holdship et al. (2016) showed H₂S exhibits the same behavior as NH₃, dropping in abundance in high-velocity gas. Given the promise of this method, molecules that exhibit this behavior are explored so this analysis can be easily extended by observing lines of these molecules. It should be noted that the key assumption of the above studies is that the decrease in emission intensity with velocity between two molecules emitting from the same source is due to a similar change in abundance with velocity. Excitation effects will complicate this, but when comparing

optically thin emission scaled to match at peak velocity, it should hold as verified with radiative transfer modeling in Viti et al. (2011).

NH₃ and H₂S are unlikely to be the only molecules enhanced in shocks and subsequently destroyed in hot gas. Species other than H₂O will remain enhanced throughout a shock and will thus be equally able to trace the whole post-shock region. H₂O can be difficult to observe with ground-based telescopes, so an alternative species showing a similar behavior could be used as a proxy. The modeling suggests two groups of molecules: shock-tracing (H₂O-like) and shock-destroyed (NH₃-like). Fractional abundances through a shock model for selected species from each categories are shown in Figure 4 to illustrate their behavior. Observing molecules from each group and comparing their velocity profiles and abundances to the predictions of the model can constrain the shock properties.

A species does not necessarily fit one category or the other for all shock conditions. Their behavior is the result of temperature and density profiles of the shock, and so depends on the shock speed and initial density. In Table 3, all modeled pairs of shock speed and initial density are shown, with lists of

shock-tracing and shock-destroyed species in blue and red, respectively. Species that exhibit more complicated behavior and cannot be easily categorized are simply omitted. Some species, such as H₂O, are almost always found to fall into one category; this allows them to be consistently used for a given purpose, i.e., H₂O can be used to trace the full shock in most cases. However, some species will demonstrate a certain behavior only in very specific conditions, and can be used to limit possible shock properties when this is observed. For example, H₂O does not trace the full shock if the initial density is 10³ cm⁻³ and the shock is low-speed. Such species are displayed in bold on Table 3.

The shock-destroyed and -tracing molecules are both enhanced by the sputtering caused by the shock. This is important because molecules that show low abundances in the pre-shock gas are unlikely to have large contributions from unshocked gas when a shocked region is observed. For example, CO shows perfect shock-tracing behavior, but is also abundant in the pre-shock gas—and its abundance is rather insensitive to shock interaction. As chemistry is the focus for this work, such species are omitted. However, note that high-J transitions of species such as CO are unlikely to be excited by ambient gas, so they can be reliably used to trace the full extent of the shock.

5. Summary

A publicly available, time-dependent, gas-grain chemical model, UCLCHEM, has been described in detail. The code solves the coupled system of ODEs that represents the chemical network when using the rate equation method of modeling chemistry. Various non-thermal and thermal desorption mechanisms are implemented along with gas phase reactions and a simple grain network. The physical models that can be used to control the gas conditions, which in turn affect the chemistry, are also described.

We present the code via two applications of the new UCLCHEM: a simple prestellar collapse model that represents a typical use of the code, and an investigation into molecular tracers for C-type shocks. This code is available at <https://uclchem.github.io>, and actively maintained and developed by research teams at UCL and Queen Mary's College London. Diffusion chemistry on the grain surfaces and improvements to the shock model implementation are in progress, and users will benefit from continued development.

J.H. is funded by an STFC studentship. I.J.-S. acknowledges financial support received from the STFC through an Ernest Rutherford Fellowship (proposal number ST/L004801/2). The

authors would like to thank D. Quenard and A. Coutens for helpful discussions that contributed to the code.

References

- Aikawa, Y., Ohashi, N., & Herbst, E. 2003, *ApJ*, **593**, 906
 Anderl, S., Guillet, V., Pineau des Forêts, G., & Flower, D. R. 2013, *A&A*, **556**, A69
 Asplund, M., Grevesse, N., Sauval, A., & Scott, P. 2009, *ARA&A*, **47**, 481
 Ayotte, P., Smith, R. S., Stevenson, K. P., et al. 2001, *JGRE*, **106**, 33387
 Bell, T. A., Hartquist, T. W., Viti, S., & Williams, D. A. 2006, *A&A*, **459**, 805
 Bisbas, T. G., Bell, T. A., Viti, S., Yates, J., & Barlow, M. J. 2012, *MNRAS*, **427**, 2100
 Bonnor, W. B. 1956, *MNRAS*, **116**, 351
 Brown, P. N., Byrne, G. D., & Hindmarsh, A. C. 1989, *SIAM Journal on Scientific and Statistical Computing*, **10**, 1038
 Burke, D. J., & Brown, W. A. 2010, *PCCP*, **12**, 5947
 Codella, C., Lefloch, B., Ceccarelli, C., et al. 2010, *A&A*, **518**, L112
 Collings, M. P., Anderson, M. A., Chen, R., et al. 2004, *MNRAS*, **354**, 1133
 de Jong, T. 1977, *A&A*, **55**, 137
 Draine, B., Roberge, W., & Dalgarno, A. 1983, *ApJ*, **264**, 485
 Flower, D. R., & Pineau des Forêts, G. 2015, *A&A*, **578**, A63
 Foster, P. N., & Chevalier, R. A. 1993, *ApJ*, **416**, 303
 Gómez-Ruiz, A. I., Codella, C., Lefloch, B., et al. 2014, *MNRAS*, **446**, 3346
 Gómez-Ruiz, A. I., Codella, C., Viti, S., et al. 2016, *MNRAS*, **462**, 2203
 Gueth, F., Guilloteau, S., Dutrey, A., & Bachiller, R. 1997, *A&A*, **323**, 943
 Guillet, V., Pineau des Forêts, G., & Jones, A. 2010, *A&A*, **527**, 123
 Gusdorf, A., Cabrit, S., Flower, D. R., & Pineau des Forêts, G. 2008, *A&A*, **482**, 809
 Holdship, J. 2017, UCLCHEM, v.1.0, Zenodo, doi:10.5281/zenodo.580044
 Holdship, J., Viti, S., Jimenez-serra, I., et al. 2016, *MNRAS*, **463**, 802
 Jimenez-serra, I., Caselli, P., Martín-Pintado, J., & Hartquist, T. 2008, *A&A*, **482**, 549
 Kandori, R., Nakajima, Y., Tamura, M., et al. 2005, *AJ*, **130**, 2166
 Lefloch, B., Vastel, C., Viti, S., et al. 2016, *MNRAS*, **462**, 3937
 Maret, S., & Bergin, E. 2015, *Astrochem: Abundances of Chemical Species in the Interstellar Medium*, Astrophysics Source Code Library, ascl:1507.010
 McElroy, D., Walsh, C., Markwick, A. J., et al. 2013, *A&A*, **550**, A36
 Occhiogrosso, A., Vasyunin, A., Herbst, E., et al. 2014, *A&A*, **564**, A123
 Rawlings, J. M. C., Hartquist, T., Menten, K., & Williams, D. A. 1992, *MNRAS*, **255**, 471
 Roberts, J. F., Rawlings, J., Viti, S., & Williams, D. A. 2007, *MNRAS*, **382**, 733
 Röllig, M., Abel, N. P., Bell, T., et al. 2007, *A&A*, **467**, 187
 Schmalzl, M., Visser, R., Walsh, C., et al. 2014, *A&A*, **572**, A81
 Suutarinen, A. N., Kristensen, L. E., Mottram, J. C., Fraser, H. J., & van Dishoeck, E. F. 2014, *MNRAS*, **440**, 1844
 Tafalla, M., Myers, P. C., Caselli, P., & Walmsley, C. M. 2004, *A&A*, **416**, 191
 Van Loo, S., Ashmore, I., Caselli, P., Falle, S. A. E. G., & Hartquist, T. W. 2009, *MNRAS*, **395**, 319
 Viti, S., Collings, M. P., Dever, J. W., McCoustra, M. R. S., & Williams, D. A. 2004, *MNRAS*, **354**, 1141
 Viti, S., Jimenez-serra, I., Yates, J. A., et al. 2011, *ApJL*, **740**, L3
 Viti, S., Roueff, E., Hartquist, T. W., des G. Pineau des Forêts, G. P., & Williams, D. A. 2001, *A&A*, **370**, 557
 Wakelam, V., Herbst, E., Loison, J.-C., et al. 2012, *ApJS*, **199**, 21
 Wilson, T. L. 1999, *RPPH*, **62**, 143