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# *IFCdiff*: A content-based automatic comparison approach for IFC files

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## Abstract

As the usage of IFC (Industry Foundation Classes) files in construction industry is on the dramatic increase, it often requires effective IFC comparison methods to keep track of important changes occurring during the lifecycle of construction projects. However, most IFC comparisons are based on a visual inspection, a manual count and a check of selective attributes. Although a few techniques about automatic IFC comparisons have been developed recently, they are usually very time-consuming, and are sensitive to the GUID change or redundant instances in IFC files. To address these issues, this paper presents a content-based automatic comparison approach, named *IFCdiff*, for detecting differences between two IFC files. This approach starts with a comprehensive analysis of the structure and content of each IFC file, and then constructs its hierarchical structure along with eliminating redundant instances. Next, the two hierarchical structures are compared with an iterative bottom-up procedure instead of the original files. The presented approach fully takes into account the content of IFC files fully without the need of flattening instances in IFC files. In contrast with previous methods, our approach can greatly reduce the computational time and space, and the comparison result is not sensitive to re-

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dundant instances in IFC files. Finally, we demonstrate a potential application to incremental backup of IFC files. The software can be found at: <http://cgcad.thss.tsinghua.edu.cn/liuyushen/ifcdiff/>.

*Keywords:* Building Information Modeling (BIM), Industry Foundation Classes (IFC), IFC comparison, Change detection, Similarity and difference

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## 1 Introduction

During the last decade, Building Information Modeling (BIM) has received a considerable amount of attention in the domain of Architecture, Engineering and Construction (AEC) to support lifecycle data sharing [1]. As an open and neutral data format specification for BIM, Industry Foundation Classes (IFC) [2] plays a crucial role to facilitate interoperability between various software platforms. The IFC data format has been widely supported by the market-leading BIM software vendors. Many recent studies also demonstrate the IFC viability in various applications, such as evaluation of design solutions [3], virtual construction [4], construction management [5], model checking [6, 7], path planning [8], file optimization [9], semantic annotation [10] and information retrieval [11].

As the usage of IFC files in construction industry is on the dramatic increase, it often requires an effective IFC comparison method to keep track of important changes occurring during the lifecycle of construction projects. The IFC comparison aims to analyze and identify the differences and similarities between two IFC files. It is a fundamental problem which may arise in many BIM-based applications, such as collaborative building design [12], incremental backup of files, construction project management [5], product data exchange [13, 14, 15], conformance checking [14], handover for operation and maintenance [14]. Previous IFC comparisons are usually based on a visual inspection, a manual count and a check of selective attributes [14, 16, 17, 18]. However, due to the large file sizes and the complex inheritance and referencing relationships of IFC files, such a way of manual inspection is often time-consuming and error-prone; furthermore, it can only report a partial and illustrative view of the compared files [13]. Although a few recent studies have been developed for automatic IFC comparison [13, 17, 19], their methods are usually very time-consuming, and are sensitive to the globally unique identifiers (GUID) change [17, 19] or redundant instances [13] within IFC files.

31 To address these issues, this paper proposes a content-based automatic  
32 IFC comparison approach, named *IFCdiff*, for tracking differences or detect-  
33 ing changes between two IFC files. Our approach starts with a comprehen-  
34 sive analysis of structure and content of each IFC file, and then constructs  
35 its hierarchical structure along with eliminating redundant instances at each  
36 level. Next the two hierarchical structures are compared with each other  
37 for detecting changes in an iterative bottom-up procedure. Our approach  
38 fully takes into account the content of IFC files and makes good use of the  
39 hierarchical structure of IFC files. Thus, our approach can greatly reduce  
40 the computational time and space, and the comparison result is not sensitive  
41 to redundant instances within IFC files. In addition, we also demonstrate a  
42 potential application of our approach to incremental backup of IFC files.

43 The paper is organized as follows. Section 2 reviews the related work and  
44 summarizes the existing problems. Section 3 introduces some basic concepts  
45 and terms of IFC files. Section 4 gives a detailed description of our approach.  
46 Section 5 demonstrates the experimental results and a potential application  
47 to incremental backup of IFC files. Finally, Section 6 concludes this paper,  
48 summarizes our contributions and discusses some future work.

## 49 **2. Related work**

50 Early studies of IFC comparison mainly conducted a visual inspection  
51 of models and a check of selective attributes in the original and exchange  
52 models [14, 16, 17, 18]. The visual inspection can be done with various IFC  
53 viewers that are available, while the attribute analysis is usually a manual  
54 check for building elements. However, only using a visual and manual way  
55 for comparing IFC files is inaccurate and incomplete due to the complex  
56 referencing and inheritance structure of IFC files [13]. The manual way is  
57 useful for only small and simple IFC models, whereas it is not practical for  
58 large and complex models in the actual construction projects. Consequently,  
59 there is an urgent need for developing automatic IFC comparison tools in the  
60 scenario of IFC-based data management.

### 61 *2.1. Plain text comparison*

62 There are various approaches in use for performing automatic comparison  
63 of one IFC file to another. An IFC file is a plain text (ASCII) format with  
64 the extension “\*.ifc”, which is specified by IFC and ISO 10303-21 [20] (also  
65 known as “STEP physical file”). Therefore, a direct approach is to use plain

66 text comparison tools for directly comparing two IFC files, regardless of  
67 information content of models. Some widely used text comparison tools [21]  
68 such as *diff*, *DiffMerge*, *cmp*, *FileMerge*, *SVN*, *CVS* and *BCompare*, can be  
69 conducted for this purpose. These tools usually compute the longest common  
70 subsequence and highlight the differences between files. However, pure text  
71 comparison does not consider specific data organization and representation  
72 of an IFC file which includes a complex referencing and inheritance structure.  
73 Therefore, the traditional text comparison tools are not suitable for IFC file  
74 comparison.

### 75 2.2. GUID-based IFC comparison

76 Another class of approaches is based on the globally unique identifier  
77 (GUID) [17, 19] which is a unique identifier for object instances across  
78 applications and systems. The GUID-based comparison criteria is as follows.  
79 If there is an instance in one IFC file which has the same GUID as an instance  
80 in another IFC file, they can be considered as the same instance; otherwise,  
81 they are considered as different even with the same attributes of the entity  
82 or of its reference entities. The GUID-based comparison is widely adopted  
83 by many commercial BIM softwares such as *Autodesk Revit*, *Navisworks* and  
84 *Graphisoft ArchiCAD*. Some research articles [19, 13] also discussed how to  
85 use the GUIDs for measuring the differences between IFC files. More recently,  
86 Oraskari et al. [22] presented RDF-based signature algorithms for computing  
87 differences of IFC models, but their algorithms is still closely related to the  
88 usage of GUIDs.

89 However, in the IFC specification, only the entities inherited from *IfcRoot*  
90 has a GUID as one of its attributes, while many other entities (e.g. *IfcProp-*  
91 *ertySingleValue* which are *IfcPropertySet*) not inherited from *IfcRoot* have  
92 no GUID [19, 13]. In addition, the GUIDs of instances are often changed  
93 during the data exchange between different systems even without any mod-  
94 ification to the model itself. Therefore, the GUID-based comparison is not  
95 a reliable approach to distinguish two IFC files even if it is quite simple and  
96 fast for comparison.

### 97 2.3. Graph-based IFC comparison

98 A third type of approaches was suggested by Arthaud and Lombardo  
99 [12] in the co-design scenario, which compares two oriented graphs generated  
100 by two IFC files. From this, it is possible to track the differences between

101 two IFC models. However, the matching process of nodes between two ori-  
102 ented graphs still complies with the GUID comparison, where the instances  
103 without GUIDs are ignored in the comparison process. Such a graph-based  
104 IFC comparison is a non-trivial and time-consuming task for large models.  
105 Furthermore, this approach does not handle duplicate data instances in IFC  
106 files. In practice, the IFC files generated by various software platforms often  
107 include a large number of duplicate data instances [9, 13], which should be  
108 processed in the process of IFC comparison. We will discuss this issue in  
109 Section 2.5.2 in detail.

#### 110 2.4. Flattening-based IFC comparison

111 The fourth type of approaches, presented by Lee et al. [13], utilizes a  
112 recursive strategy to flatten the instances in two IFC files, and then compares  
113 the flattened data instances instead of the original ones. The “flattening”  
114 process is to replace all the reference numbers with their actual values in each  
115 IFC file, which makes an IFC file into a structure that does not include any  
116 referencing or inheritance structure [13]. This overcomes the difference of  
117 reference numbers included in attribute values when comparing pairs of data  
118 instances. As a result, IFC comparison is simplified to pure string comparison  
119 after flattening.

120 This flattening-based method firstly reads two IFC files and parses data  
121 into instance name, entity name, and attribute values before comparing. In  
122 the following example of one data instance, #90 is the instance name, IFC-  
123 SLAB is the entity name, and the remaining information within parentheses  
124 is the attribute values.

125  
126 

#90=IFCSLAB(2VLPPLMIR7fBUKZN0XN2MZ, #13, S- LAB_006, , \$, #335, #320, \$, .FLOOR.)
--

  
127

128 Since different BIM modeling systems might export IFC files in various ways.  
129 As a result, the instance names and reference numbers might be different.  
130 To overcome this difference in referencing mechanisms, the files should be  
131 “flattened” first, i.e., making files in a structure that does not include any  
132 referencing or inheritance structure by replacing the reference identifier num-  
133 bers with their actual attribute values. The following shows the flattened  
134 data instance of #90.

135

```

136 #90=IFCSLAB (2VLPPLMIR7VLPPLMIR7fBUKZN0XN2MZ,
    $, UNDEFINED, $, $, $, $, $, $, $, ORGANIZATIONNAME,
    $, $, $, $, GS, GRAPHISOFT, GRAPHISOFT, $, $, 9.0, A-
    CAD9.0, ARCHICAD, $, .NOCHANGE., $, $, $, 1149148841, S-
137 LAB_006, , $, $, (0.,0.,0.), (0.,0.,1.), (1.,0.,0.), (0.,0.,0.), (0.,0.,1.),
    (1.,0.,0.), (-43500.,14500.,-200.), (0.,0.,1.), IFCPARAMETERVAL-
    UE(0.)), ((0.,0.), IFCPARAMETERVALUE(90.)), .T., .CARTE-
    SIAN.), .F., (0.,0.,0.), (0.,0.,1.), (1.,0.,0.), (0.,0.,1.), 200.)), $,
    .FLOOR.)

```

138 Such a flattening process overcomes the difference of reference numbers in-  
139 cluded in attribute values when comparing pairs of data instances. As a  
140 result, IFC comparison is simplified to pure string comparison after flatten-  
141 ing.

142 The process of file comparison in [13] consists of three main steps: (1)  
143 first parsing all data instances and flattening them, and then (2) comparing  
144 the flattened instances while ignoring their GUIDs, finally (3) computing the  
145 similarity. One main advantage of the flattening-based comparison approach  
146 is that it is insensitive to the change of GUIDs of data instances in IFC files.  
147 However, the flattening-based file comparison is usually time-consuming for  
148 large models, and it is also sensitive to redundant instances appearing in IFC  
149 files. In addition, this approach does not deal with the order changes of the  
150 properties in property sets in data instances. For example, an data instance  
151 *IfcPropertySet* is given below.

```

152 #145=IFCPROPERTYSET('3wesF7dHX9B9kkD2hgAhST', #33,
153 'PSet_Revit', $, (#133, #134, #135, #136, #137, #138));
154

```

155 In the instance #145, the last attribute is a collection of attribute instances,  
156 i.e. (#133, #134, #135, #136, #137, #138), in which each attribute  
157 instance (e.g. #133) is an *IfcPropertySingleValue* indicating an attribute  
158 value. In this collection, the order of these attribute instances might change  
159 during data exchange between different BIM software. Therefore, this re-  
160 quires a special treatment in the file comparison process. However, the  
161 flattening-based IFC comparison does not consider this situation, which may  
162 result in that the same data instances but with different orders of attributes  
163 are considered to be different.

164 *2.5. Summarizing the existing problems*

165 After reviewing the existing approaches [12, 13, 17, 19], we summarize  
166 the existing problems as follows.

167 *2.5.1. Sensitivity for GUID changes*

168 Although the GUID-based approach [17, 19] is simple and fast without  
169 comparing all attribute values, the GUIDs of data instances are often changed  
170 in data exchange from different systems. Therefore, it is not an appropriate  
171 way for identifying the differences between IFC files. The graph-based com-  
172 parison [12] is time-consuming for large models, and it still complies with the  
173 GUID comparison during node matching. In addition, this approach is also  
174 sensitive to the redundant instances.

175 In contrast, our method compares the contents and structures of IFC files  
176 through an iterative procedure, which does not rely on GUIDs.

177 *2.5.2. Sensitivity for redundant instances*

178 Redundancy in information theory is the number of bits used to transmit  
179 a message minus the number of bits of actual information in the message.  
180 Informally, it is the amount of wasted “space” used to transmit certain data  
181 [9]. Many previous studies (e.g. [9, 13, 16, 19]) have introduced that the  
182 exported IFC files in practice often contain a large number of redundant in-  
183 formation. Our recent paper [9] also illustrated several possible reasons for an  
184 abundance of redundancy in the exported IFC files. For instance, *differences*  
185 *of model mapping mechanism* between various BIM software platforms and  
186 the standard IFC data may produce a great deal of redundancy, and *various*  
187 *possibilities offered by the IFC specification* can cause redundancy too [9].

188 One typical example of redundancy in IFC files is the *identical data in-*  
189 *stances* [9, 13], which are roughly defined as multiple instances of the same  
190 entity with the same entity name and attribute values, but possibly with  
191 different instance names. For example, the duplicate instances of the *Ifc-*  
192 *CartesianPoint* entity with the same value are one common example of re-  
193 dundant information. The identical data instances are the representative of  
194 the redundancy that should be dealt with in the process of IFC comparison.  
195 Complying with information theory, our approach eliminates the problem of  
196 redundancy existing in IFC files before comparing in order to remove the  
197 influence on the similarity caused by the redundant instances.

198 The metric computation based on flattening the instances in [13] is sensi-  
199 tive to redundant instances in IFC files. In Eq. (1), the number of matching



200 instances is highly relevant to the number of redundant instances in the IFC  
201 files. Assuming that there are a large number of duplicate instances in File  
202 A matched to data in File B, the similarity rate in Eq. (1) will be very high  
203 even close to 100%. This is unreasonable, because various IFC files (with or  
204 without duplicate instances) of the same building model should describe the  
205 same data model. A robust similarity rate computation should be insensitive  
206 to the number of redundant instances in IFC files. Although Lee et al. [13]  
207 also presented the *matching rate* for indicating how often instances in File  
208 A are redundantly produced in File B, it cannot improve the similarity rate  
209 computation in essence.

210 Being different with the flattening-based comparison approach, our ap-  
211 proach constructs the hierarchical structures of IFC files along with eliminat-  
212 ing redundant instances. Then the two hierarchical structures are compared  
213 with an iterative bottom-up procedure instead of the original IFC files. By  
214 removing the redundant instances while keeping the complete IFC model-  
215 s, the approach can overcome the influence arising by redundant instances  
216 in IFC files. Consequently, our approach can obtain a stable and reliable  
217 similarity rate compared with the flattening-based approach [13].

### 218 2.5.3. Time-consuming to calculation

219 The flattening-based approach [13] is also time-consuming for comparison  
220 of large IFC files. On the one hand, the comparison between a large number  
221 of duplicate instances existing in IFC files will take a lot of time; while it is  
222 in fact unnecessary. On the other hand, after all instances in an IFC file are  
223 flattened, the generated strings of flattened instances become quite long due  
224 to the complex referencing and inheritance structure of IFC. It will cost a lot  
225 of time and space to complete the process of instance matching. In general,  
226 the flattening process will increase the size of an IFC file several times or even  
227 dozens of times. For example, a 10M IFC file in our test cases is increased  
228 to 70M after flattening.

229 Compared with the flattening-based approach in [13], our approach avoids  
230 the procedure of flattening instances and is able to gain the similarity rate  
231 in a much shorter time. Furthermore, since the redundant instances are  
232 removed from the original files when using our approach, the number of data  
233 instances to be compared is decreased significantly. This greatly improves  
234 the comparison efficiency.

235 A more formal investigation is given in Appendix A for discussing the  
236 complexities of the mentioned algorithms.

#### 237 2.5.4. Other issues

238 In the previous work, the order problem of aggregation attributes was not  
239 considered in the process of IFC comparison. In IFC, a lot of attributes are in  
240 the form of a collection of reference numbers. For example, the relationship  
241 object associates one object with several other objects or attributes, and these  
242 objects or attributes are recorded as reference numbers in a set. Another  
243 example is the property set which includes some reference numbers and each  
244 of them stands for one property. Since different systems export data in  
245 different ways, the order of aggregation attributes might change during data  
246 exchange. However, the previous approaches (also including flattening-based  
247 approach [13]) regard this case, i.e. that those instances with the same  
248 attribute sets but in different order, as different instances. In contrast, our  
249 approach handles the order problem of attributes, which produces a stable  
250 similarity rate.

251 The GUID-based and flattening-based approaches mainly focus on textu-  
252 al comparison between two IFC files. However, since the readability of IFC  
253 text file is poor, it is non-trivial for users to find the differences between geo-  
254 metric models only through text comparison. In fact, each IFC file includes  
255 geometric information which represents a 3D building model. If the textual  
256 comparison results can be associated with the 3D model, it will enable users  
257 to intuitively understand the differences and changes between models. This  
258 paper develops a prototype *IFCdiff* viewer specifically designed to highlight  
259 the different geometric objects between models.

#### 260 2.6. Tree compression

261 In computer science, tree compression (or named tree compaction) is a  
262 common task and well-studied. Given a tree, the task is to map it as com-  
263 pactly as possible to memory [23], where the range of the mapping depends on  
264 specific applications. Many methods such as arithmetic coding and Huffman  
265 coding can be used for encoding and decoding of trees on data compression.  
266 There have been some typical applications of the tree compression methods  
267 such as the compression of pixel trees, syntax compression of program files,  
268 and the compression of XML document trees [24].

269 In this paper, we simplify each IFC file as a tree structure and remove  
270 the redundant data instances from this tree, which can be regarded as an  
271 application of tree compression to IFC files. Then, the compressed tree  
272 structures derived from two IFC files are compared instead of comparing the  
273 original IFC files. When the compression of tree structures is considered,

274 two objectives are often involved. The first objective is to reduce the space  
275 needed for storing a tree itself, and the second one is to reduce the operation  
276 time on the specific application. Our method meets both of the requirements  
277 because the space can be saved through removing the redundant nodes while  
278 accelerating the functionality of the operations (i.e. IFC comparison).

### 279 3. Basic terms and IFC hierarchical structure

280 This section introduces some basic terms used in this paper and IFC  
281 hierarchical structures.

#### 282 3.1. Basic terms used in the IFC file

283 As the ISO 16739 standard, IFC defines a conceptual data schema and an  
284 exchange file format of building information models. An IFC data file is in an  
285 ASCII text format with the extension “\*.ifc”, which uses the STEP physical  
286 file structure according to ISO 10303-21 [20]. The IFC file is composed of  
287 a *header section* and a *data section* [25], as shown in Figure 1. The header  
288 section describes basic information including the file description, the date  
289 and time, the schema version, etc. The data section defines the BIM data  
290 including a large number of *entity instances* (or named *data instances*). Each  
291 entity instance takes “#” as the beginning of the sentence and has *instance*  
292 *name*, *entity name* and a list of *attribute values*. The instance name (e.g.  
293 “#3967”) is unique within the scope of an IFC file, which can also be used as  
294 a *reference id* cited by other entity instances. An example IFC file is shown  
295 in Figure 1, where some basic terms are illustrated.

296 Note that the instance names in two IFC files are independent of each  
297 other, so they cannot be used as a feature to distinguish two data instances.  
298 In our approach, the entity name and attribute values are considered for  
299 instance comparison.

#### 300 3.2. Hierarchical structure of the IFC file

301 IFC divides all entities into *rooted* and *non-rooted* entities. Rooted enti-  
302 ties derive from the most abstract class *IfcRoot* and each one has a GUID  
303 along with attributes. Non-rooted entities have no GUID, and data instances  
304 only exist if referenced from a rooted data instance directly or indirectly.

305 The IFC data model is essentially constructed in a hierarchical structure,  
306 generally with the rooted entity *IfcProject* as the *root node*. This structure  
307 is named *IFC hierarchical structure* in this paper. The data instances (e.g.

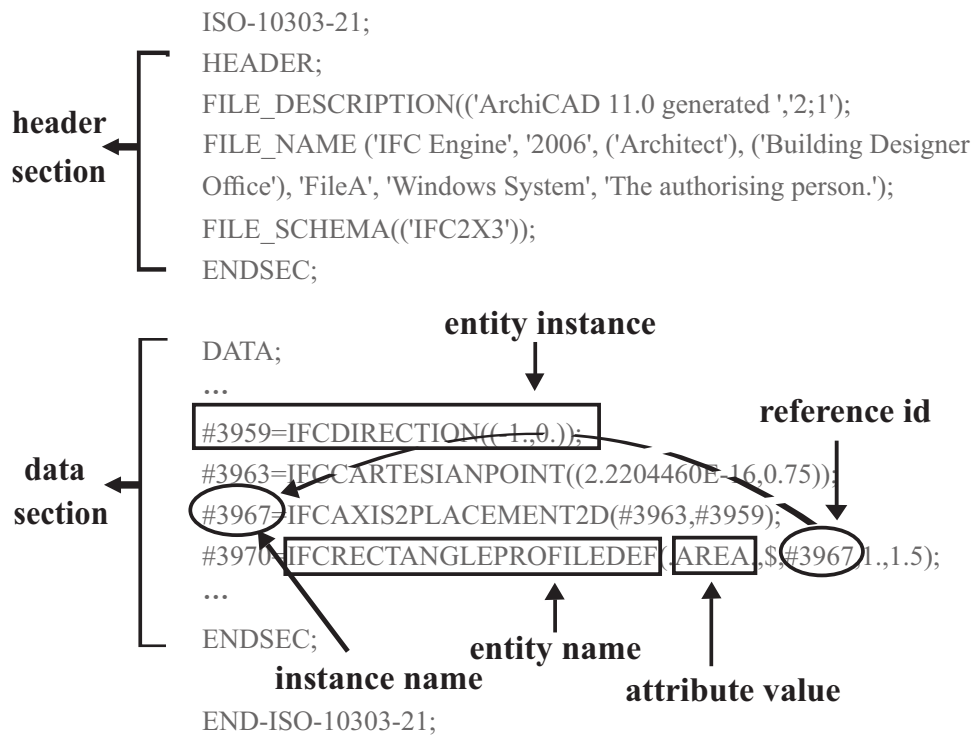


Figure 1: The basic terms used in the IFC file.

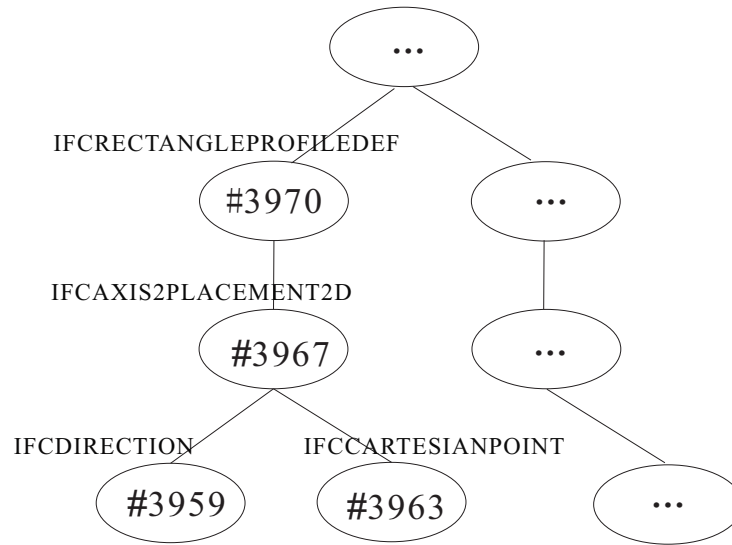


Figure 2: A partial IFC hierarchical structure corresponding to the file fragment in Figure 1, where each node also indicates its corresponding entity name.

308 non-rooted entities *IfcDirection* and *IfcCartesianPoint*) that do not include  
 309 any reference id in their attributes are considered as the *terminal nodes*, or  
 310 level 0. The data instances that directly cite the level 0 nodes are their *parent*  
 311 *nodes*, or level 1. Consequently, the data instances are structured as the level  
 312  $n$  nodes, if they are the parent nodes of level  $n - 1$ . The similar hierarchi-  
 313 cal representation of IFC file was also used in other IFC-based applications  
 314 including IFC compression [9] and partial model extraction [26].

315 Figure 2 shows a partial IFC hierarchical structure corresponding to the  
 316 file fragment in Figure 1. In Figure 2, the data instances (e.g. “#3959” and  
 317 “#3963”) are recognized as the terminal nodes (i.e. level 0), whose parent  
 318 node is the data instance “#3967” (i.e. level 1). The data instance “#3970”  
 319 (i.e. the parent node of “#3967”) is recognized as the level 2.

#### 320 4. The content-based IFC comparison approach

321 To achieve a fast and redundancy-insensitive IFC comparison, this sec-  
 322 tion introduces a content-based automatic comparison approach for detecting  
 323 changes between two IFC files. By analyzing the content of each IFC file,  
 324 the approach first constructs the IFC hierarchical structure along with elim-  
 325 inating duplicate data instances. Then these two hierarchical structures are

326 compared with an iterative bottom-up procedure. The main procedure of  
327 our approach is illustrated in Figure 3. Starting with two IFC files as input,  
328 our approach contains four steps as follows.

329 Step 1: Preprocess the data instances and construct the IFC hierarchical  
330 structures (see Section 4.1). This step first removes redundant infor-  
331 mation in each data instance, and then extracts three basic terms(i.e.  
332 instance name, entity name and attribute values) from each data in-  
333 stances. Next, based on the extracted terms and their referencing re-  
334 lationships, the hierarchical structures of two IFC files are constructed  
335 for further comparison.

336 Step 2: Compare the terminal nodes between two IFC hierarchical structures  
337 along with removing redundant instances (see Section 4.2). This step  
338 first identifies and groups identical data instances in the terminal nodes.  
339 Then only one data instance of each group is kept, while all other  
340 duplicate instances are removed from this group. Next, we compare  
341 the updated terminal nodes between two hierarchical structures and  
342 find the matching instances between them.

343 Step 3: Repeat Step 2 for iterative and level-by-level comparison for the  
344 remaining data instances between two files (see Section 4.3). Step 3 is  
345 a recursive and iterative process terminated until the comparing nodes  
346 reach the root node in any one of two files. The matching instances  
347 between two files are recorded in a hash table.

348 Step 4: Finally, compute the similarity rate between two IFC files (see Sec-  
349 tion 4.4). The similarity rate is defined as the rate of the number of  
350 matching instances between two files divided by the total number of  
351 instances in the target file. In addition, all matching instances between  
352 two files are saved in the hash table, and the differences between files  
353 are also recorded for further applications (e.g. incremental backup of  
354 IFC files in Section 5.6).

355 For the reader’s convenience, the target file and the source file will be  
356 referred to as **File A** and **File B** in this paper, respectively.

#### 357 *4.1. Step 1: Preprocess data instances and construct the IFC hierarchical* 358 *structures*

359 We preprocess data instances within each input IFC file, and then con-  
360 struct the IFC hierarchical structures. The preprocessing will remove redun-  
361 dant information (e.g. blank spaces and multi-lines) from each data instance

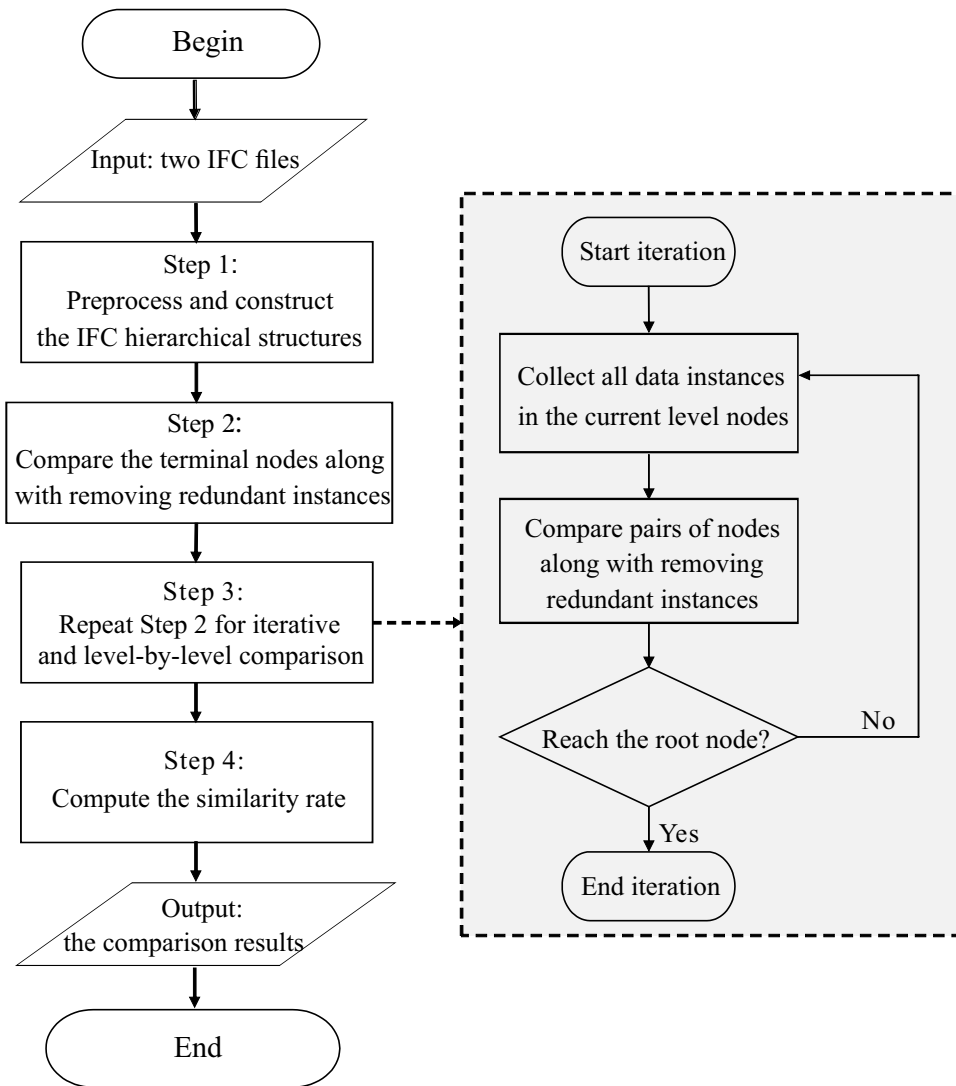


Figure 3: The flow diagram of our content-based IFC comparison approach.

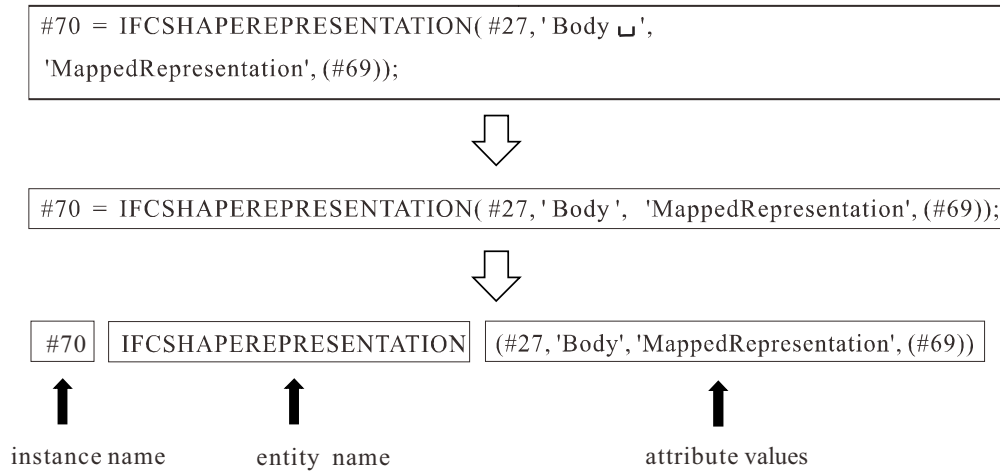


Figure 4: Illustration of preprocessing data instances within each IFC file, where three basic items will be extracted from each data instance.

362 within each IFC file. Especially, the data instance with multi-lines is convert-  
 363 ed into a single line. Then, the contents of three basic terms (i.e. instance  
 364 name, entity name and attribute values) are extracted from each data in-  
 365 stance, as shown in Figure 4. The above preprocessing is similar to the  
 366 strategy in [9]. Finally, based on the extracted terms and their referencing  
 367 relationships, the hierarchical structure of each IFC file (mentioned in Section  
 368 3.2) is constructed for further comparison.

369 Figure 5 shows an example of two file fragments from the target file  
 370 (File A) and the source file (File B), respectively. The corresponding IFC  
 371 hierarchical structures are displayed in Figure 6.

372 *4.2. Step 2: Compare the terminal nodes along with removing redundant*  
 373 *instances*

374 The constructed IFC hierarchical structure is a tree-like data structure,  
 375 so the comparison of two structures can be conducted in a similar way of  
 376 level-by-level comparison of two trees. To accomplish this goal, we need to  
 377 traverse two trees simultaneously, where this traversal visits the nodes by  
 378 levels from bottom to top. The second step of our approach is to compare  
 379 the terminal nodes between two hierarchical structures along with removing  
 380 redundant instances.

381 Firstly, for each file, we collect all data instances on the terminal nodes  
 382 which do not include any reference id in their attribute values. For example,



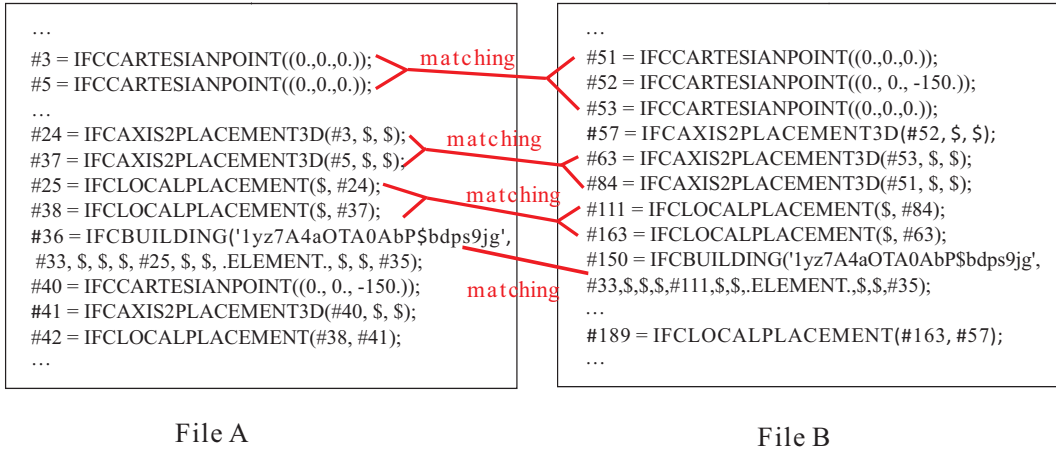


Figure 5: Illustration of two file fragments from the target file (File A) and the source file (File B), respectively. The matching instances between two file fragments are highlighted.

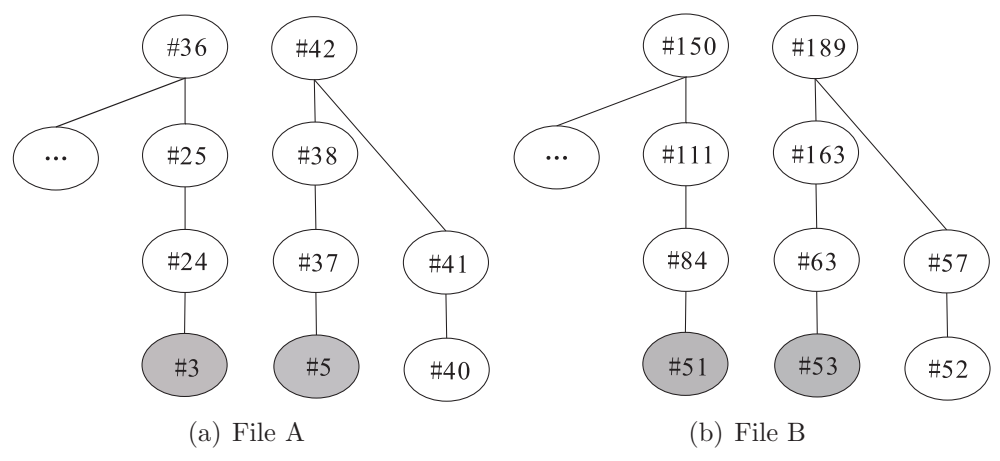


Figure 6: The two partial IFC hierarchical structures are displayed, which correspond to File A and File B (in Figure 5), respectively. Here the identical data instances (light gray nodes) on the terminal nodes (i.e. level 0) in each file are identified and grouped.

383 in File A, the data instances (#3, #5 and #40) are recognized as the terminal  
384 nodes in the hierarchical structure (see Figure 6(a)). In File B, the data  
385 instances (#51, #53 and #52) are the terminal nodes (see Figure 6(b)).

386 Secondly, for each file, we identify and group the identical data instances  
387 in the terminal nodes by comparing their entity names and attribute values,  
388 where the terminal nodes with the same value are clustered into one group.  
389 Consequently, we can obtain multiple groups of identical data instances. In  
390 Figure 6, the light grey nodes (#3 and #5) denote one group of identical  
391 data instances in File A, while #51 and #53 are grouped together in File B.

392 Thirdly, only one data instance of each group is kept, while all other  
393 duplicate instances are removed from this group. Meanwhile, the reference  
394 id of attribute values in the remaining data instances in the upper levels will  
395 be updated accordingly. As shown in Figure 7, the data instance #3 in the  
396 group is kept in File A, while #5 is deleted. Meanwhile, their upper parent  
397 nodes (i.e. #24 and #37) are respectively relocated to the data instance #3.  
398 File B is processed similarly, where #51 is kept and #53 is removed. After  
399 achieving the above process, we can remove all redundant data instances  
400 from the terminal nodes both in File A and in File B.

401 Finally, we compare the new terminal nodes (without duplicate instances)  
402 between the two hierarchical structures while ignoring the GUIDs, and find  
403 the matching instances between them. Since the data instances in the ter-  
404 minal nodes do not include any reference id, we only compare the values  
405 of data instances (i.e. their entity names and attribute values) in terms of  
406 string comparison. In Figure 7, as for the terminal nodes, #3 in File A is  
407 matched to #51 in File B, where #3 and #51 are the same as “IFCCARTE-  
408 SIANPOINT((0.,0.,0.))” by checking the original file fragments in Figure 5.  
409 In addition, we record the pair of matching instances (#3, #51) in a *hash*  
410 *table* (denoted by  $T$ ) for the upper level comparison.

#### 411 4.3. Step 3: Repeat the iterative comparison process

412 In a similar way to Step 2, we need to traverse all nodes of two IFC hierar-  
413 chical structures by levels from bottom to top. Therefore, we make use of an  
414 iterative strategy for comparing the nodes of each level along with removing  
415 duplicate instances. The procedure of iterative comparison is described as  
416 follows.

417 Firstly, for each IFC file, these data instances which directly cite the  
418 terminal nodes mentioned in Section 4.2 are collected, which will be treated  
419 as the new terminal nodes instead of the previous ones. As shown in Figure

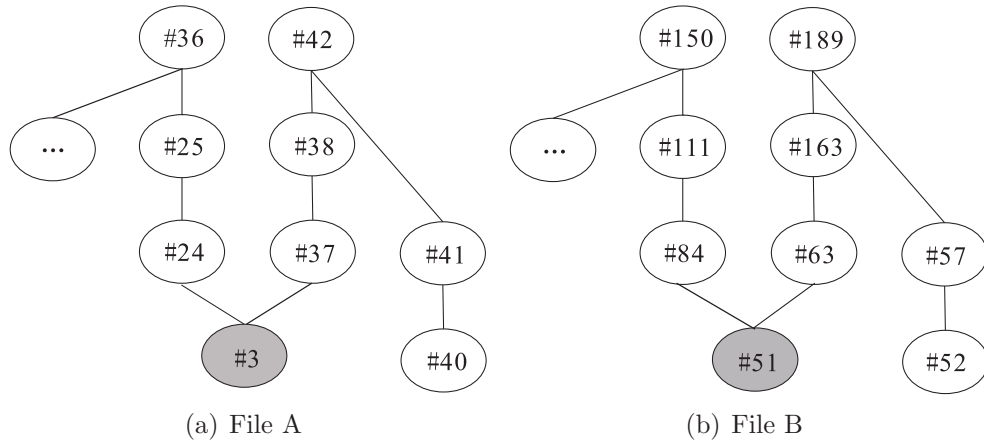


Figure 7: Remove the duplicate instances from the terminal nodes (i.e. level 0) in each file, and update the reference id of their upper parent nodes. (a) #3 is kept in File A, while #5 is removed. (b) #51 is kept in File B, while #53 is deleted.

420 8, the data instances (#24, #37 and #41) on the level 1 in File A become the  
 421 new terminal nodes instead of the previous terminal nodes (#3 and #40),  
 422 while the data instances (#84, #63 and #57) in File B are regarded as the  
 423 new terminal nodes instead of #51 and #52.

424 Secondly, in a similar way to Step 2, we group the identical data instances  
 425 in the new terminal nodes (i.e. level 1), and then the duplicate instances are  
 426 removed from this level in each IFC file. In Figure 8, the light gray nodes  
 427 (#24 and #37) denote one group of identical data instances in File A, while  
 428 the light gray nodes (#84 and #63) are another group in File B. After  
 429 removing the duplicate instances, the data instance #24 is kept in File A,  
 430 while #84 is kept in File B, as shown in Figure 9. Meanwhile, the upper  
 431 parent nodes (i.e. level 2) are updated accordingly (see Figure 9), where the  
 432 data instances (#25 and #38) are relinked to #24 in File A and the data  
 433 instances (#111 and #163) are relinked to #84 in File B.

434 Thirdly, we compare the new terminal nodes (i.e. level 1) between two  
 435 hierarchical structures while ignoring the GUIDs, and find the matching in-  
 436 stances between them. Since the data instances in the level 1 include the  
 437 reference id in their attribute values, we first compare the reference id and  
 438 then compare the remaining properties between the two data instances. In  
 439 Figure 9, the data instance #24 in File A and #84 in File B are accordingly  
 440 updated as follows.

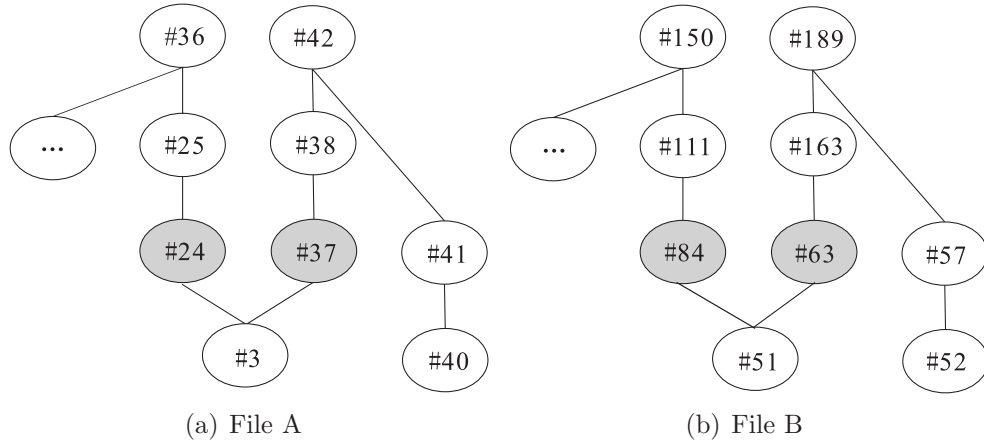


Figure 8: Illustration of the iterative comparison process on the level 1. Here we group the identical data instances (light gray nodes) in the level 1 in each file. (a) The nodes (#24 and #37) denote one group in File A. (b) The nodes (#84 and #63) are in one group in File B.

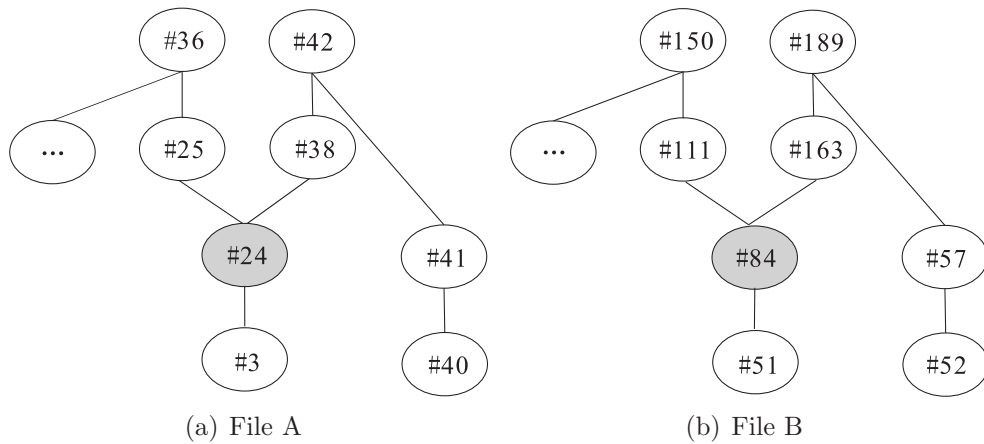


Figure 9: Illustration of the iterative comparison process on the level 1. Here we remove the duplicate instances from the nodes of level 1 in each file, and update the reference id of their upper parent nodes. (a) #24 is kept in File A, while #37 is removed. (b) #84 is kept in File B, while #63 is deleted.

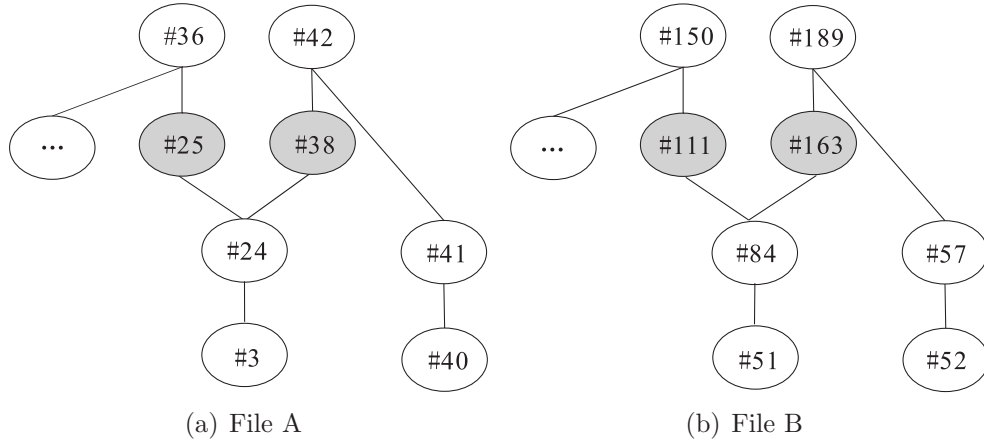


Figure 10: Illustration of the iterative comparison process on the level 2. Here we group the identical data instances (light gray nodes) in the level 2 in each file. (a) The nodes (#25 and #38) denote one group in File A. (b) The nodes (#111 and #163) denote another group in File B.

441

```
#24=IFCAXIS2PLACEMENT3D(#3, $, $);
```

442

443

```
#84=IFCAXIS2PLACEMENT3D(#51, $, $);
```

444

445

446

Since we have recorded the matching instances (#3, #51) in the hash table  $T$  in the level 0 (see Section 4.2), the reference id of #24 is the same as the one of #84. In addition, the entity names and other properties between #24 and #84 are the same, so they are a pair of matching instances in the level 1. Meanwhile, this pair of matching instances (#24, #84) are continuously added into the hash table  $T$  for the upper level comparison.

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We repeat the above procedure for further comparing the remaining data instances in two files. If the comparing nodes reach the root node (i.e. *IfcProject*) of File A or File B, the iterative comparison procedure is terminated. Figure 10 and Figure 11 show the above comparison procedure in the level 2. Figure 12 illustrates the pair of matching instances (#36, #150) in the level 3.

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#### 4.4. Step 4: Compute the similarity metric

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459

The last step of our approach is to compute the similarity rate between two IFC files. Being similar to the similarity metric used in [13], we define the

460

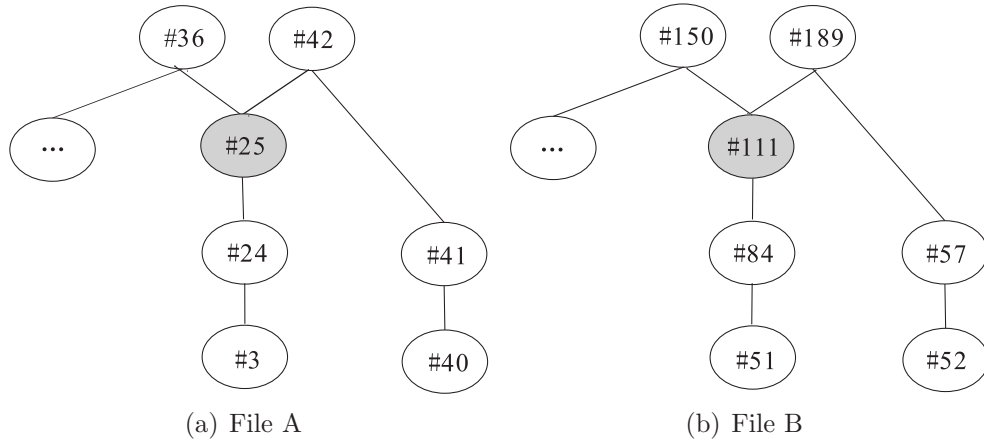


Figure 11: Illustration of the iterative comparison process on the level 2. Here we remove the duplicate instances from the nodes of level 2 in each file and update the reference id of their upper parent nodes. (a) #25 is kept in File A, while #38 is removed. (b) #111 is kept in File B, while #163 is deleted.

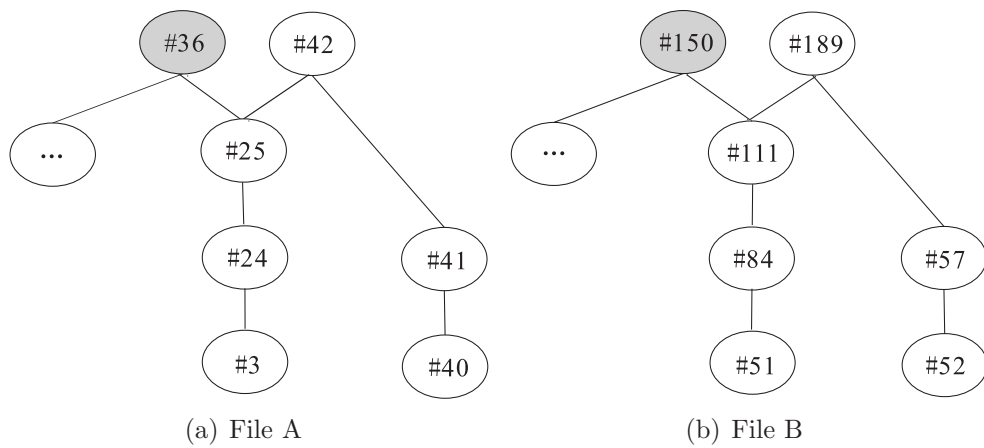


Figure 12: Illustration of the iterative comparison process on the level 3. On the level 3, #36 in File A and #150 in File B are a pair of matching instances.

461 similarity rate from File A to File B as the rate of the number of matching  
 462 instances between two files divided by the total number of instances in File  
 463 A.

$$\text{Similarity}(A, B)(\%) = \frac{|A \cap B|}{|A|}, \quad (1)$$

464 where  $|A|$  is the total number of instances in File A after removing redundant  
 465 instances, and  $|A \cap B|$  is the number of matching instances between File A  
 466 and File B along with removing redundant instances using our approach. In  
 467 contrast with the previous flattening-based approach in [13], when using our  
 468 approach, the number of matching instances in File A compared to File B is  
 469 the same with those in File B compared to File A, i.e.  $|A \cap B| = |B \cap A|$ ,  
 470 even if the input files include redundant data instances. Consequently, our  
 471 approach can obtain a stable and reliable similarity rate.

472 As for the example of two file fragments in Figure 5 used in this section,  
 473 the similarity rate from File A to File B [13] is 70.0% (7/10) based on the  
 474 flattening-based approach, while the similarity rate is 57.1% (4/7) with our  
 475 approach.

476 Finally, all matching instances between two files are saved in the hash  
 477 table  $T$ , and the differences between them are also recorded for further ap-  
 478 plications (e.g. incremental backup of IFC files in Section 5.6).

#### 479 *Computational complexity*

480 Let  $n$  and  $m$  ( $n \geq m$ ) be the number of data instances in File A and File  
 481 B, respectively. First, as for IFC hierarchical structures, it takes  $O(n)$  (for  
 482 File A) and  $O(m)$  (for File B) to preprocess all the data instances in Step  
 483 1. Meanwhile, it takes  $O(n \log(n))$  (for File A) and  $O(m \log(m))$  (for File  
 484 B) to remove the redundant data instances and update the reference id [9].  
 485 Finally, it takes  $O(n \log(m))$  to compare pairs of data instances between two  
 486 files with an iterative bottom-up procedure. As a result, the total complexity  
 487 is about  $2O(n) + 2O(n \log(n)) + O(n \log(m))$ , and therefore an upper bound  
 488 of running time is  $O(n \log(n))$ . A more detailed analysis for computational  
 489 complexity is dependent on the two IFC hierarchical structures, and we leave  
 490 it to the future work. In our implementation, we use the hash table to save  
 491 the matching instances to accelerate the node searching and comparison.

#### 492 *4.5. Improvements of approach implementation*

493 In order to address several issues mentioned in Section 2.5, we make some  
 494 improvements for the presented approach.

495 (1) *Ignoring the GUID change*

496 During data exchange, initial GUIDs of data instances often get lost or  
497 changed. Therefore, we compare pairs of data instances while ignoring their  
498 GUIDs in Step 2 and Step 3 of our approach. This can overcome the effects  
499 of GUID changes during data exchange.

500 (2) *Ignoring the change of owner history information*

501 The owner history information (*IfcOwnerHistory*) contains information  
502 about the author, create time, modeling software and so on. This information  
503 will be changed whenever an IFC file is imported and exported from a system,  
504 even if there is no change in the model itself. Therefore, to identify the actual  
505 changes between two models, the owner history information is ignored in the  
506 comparison of instance attribute values.

507 (3) *Ignoring the order change of property set*

508 The previous comparison approach does not deal with the problem of the  
509 order changes of the properties in property sets. As mentioned above, the at-  
510 tribute of *IfcPropertySet* may be a collection of some attribute instances, which  
511 requires special treatment in the file comparison process. When comparing  
512 two *IfcPropertySet* instances, we compare all attribute instances in two col-  
513 lections and find the matching data instance with the help of the hash table  
514  $T$ .

## 515 5. Experimental results and discussions

516 Our approach has been implemented in a content-based IFC comparison  
517 tool, called *IFCdiff*, with Visual C++ under Windows 8. All the experiments  
518 were run on an Intel Pentium(R) Dual-Core 3.06GHZ processor with 6 GB  
519 memory. Figure 13 shows the screenshot of the *IFCdiff* tool. The user  
520 should first open two candidate IFC files (the target and source files), and  
521 then click the button “**Compare**” to perform the comparison procedure. The  
522 comparison results and the similarity metrics are displayed at the bottom.  
523 Alternatively, one can select multiple checkboxes to ignore the GUIDs, owner  
524 history information and the order of property set.

525 In order to visualize the compared models and their differences, we also  
526 developed an *IFCdiff* viewer, as shown in Figure 14. In the main interface  
527 of the viewer in Figure 14(a), the corresponding differences of two input IFC  
528 files are highlighted in the text boxes in the middle, the similarity metrics



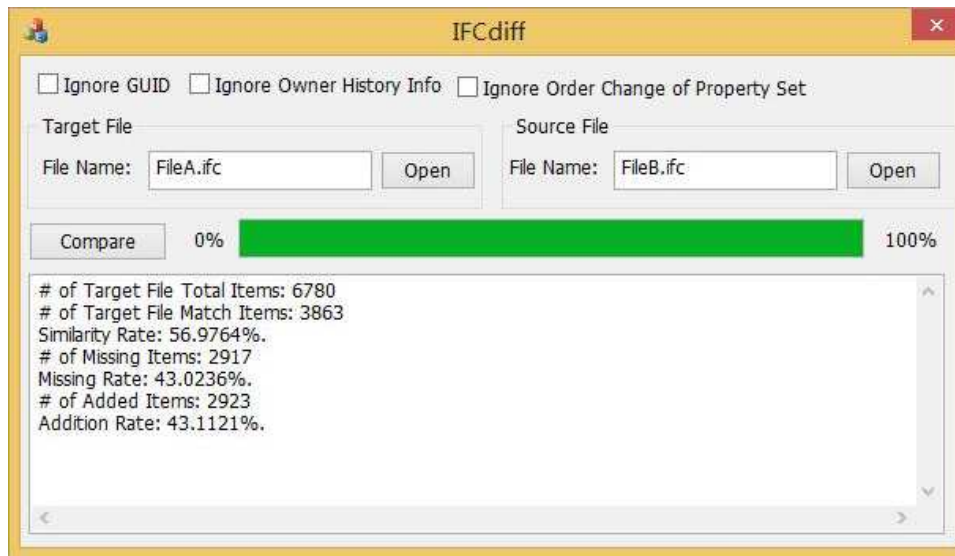


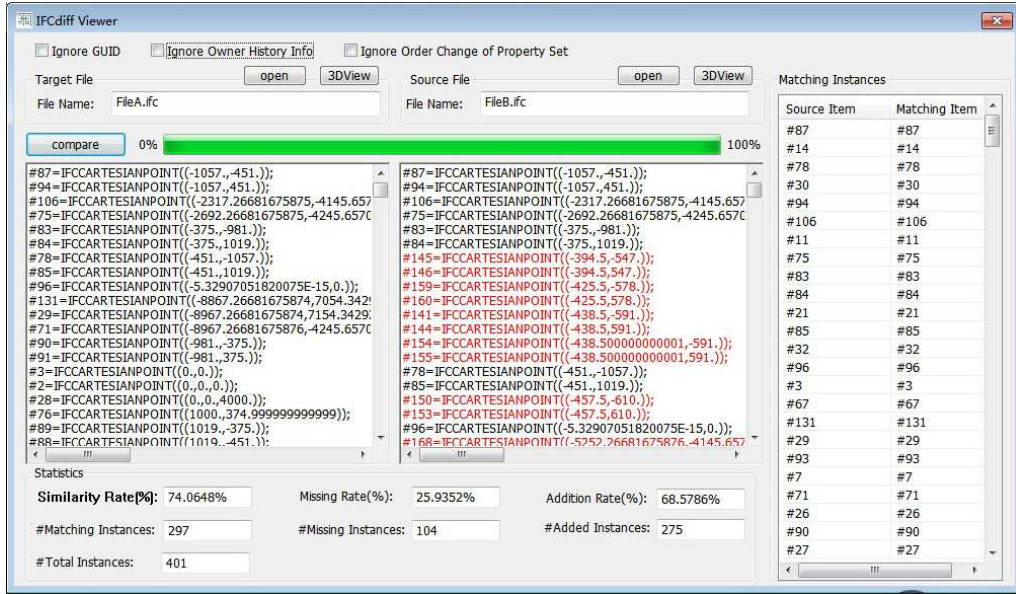
Figure 13: The screenshot of our comparison tool *IFCdiff*.

529 and a summary of the analysis are given at the bottom, and the matching  
 530 data entities between two files are listed on the right. By clicking the button  
 531 “**3DView**” of each file in Figure 14(a), the viewers of 3D models will pop  
 532 up in Figure 14(b) and Figure 14(c), where the matching building elements  
 533 are highlighted with the same color. This enables users to check the visual  
 534 differences and changes between IFC models quickly.

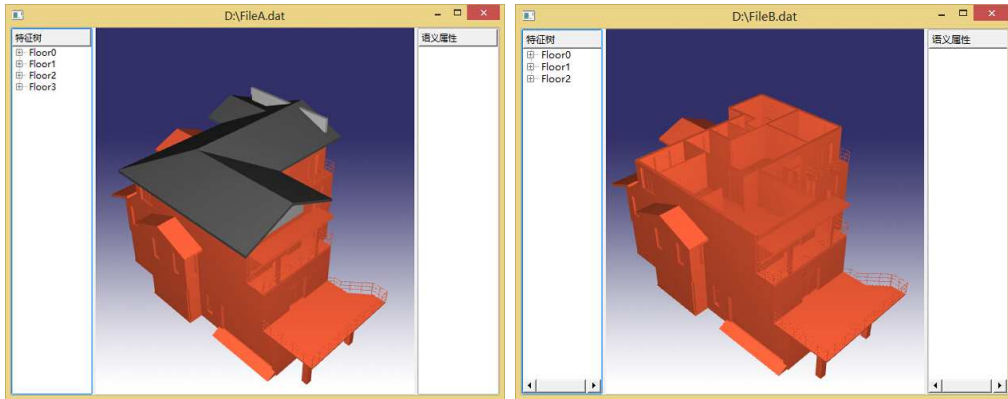
535 To evaluate the performance of the presented approach, this section tests  
 536 our approach on some selected IFC files, and the experiments are conducted  
 537 by comparing with other existing approaches. Finally, we demonstrate a  
 538 potential application to incremental backup of IFC files.

### 539 5.1. Comparison with plain text comparison methods

540 The first experiment compares our method with plain text comparison  
 541 methods. Many plain text comparison tools [21] are able to achieve file  
 542 comparison to highlight the differences between files. Such tools generally  
 543 perform string comparison of string-by-string or line-by-line and highlight the  
 544 differences and changes between two files. Here we typically choose the tool  
 545 *DiffMerge* to compare two IFC files and show their differences. In the test  
 546 case, a building model first was built in *Graphisoft ArchiCAD 16* and was  
 547 exported as File A (Figure 15(a)). Then the same model was slightly modified



(a) The main interface of the *IFCdiff* viewer



(b) 3D model of File A

(c) 3D model of File B

Figure 14: The screenshot of the *IFCdiff* viewer.

```

561 #724= IFCPROPERTYSINGLEVALUE('IsExternal', $, IFCBOOLEAN(.T.), $);
562 #725= IFCPROPERTYSET('3vCh$0ythZzhLClJPLrosJ', #15, 'Pset_WallCommon', $, (
563 #721, #722, #723, #724));
564 #727= IFCRELEDEFINESBYPROPERTIES('260N_OFnjMqInhadQjXR_Y', #15, $, $, (#509),
565 #725);
566 #731= IFCPROPERTYSINGLEVALUE('\S\6\S\(\S\N\S\;\S\O\S\_', $, IFCINTEGER(0),
567 $);
568 #732= IFCPROPERTYSINGLEVALUE('\S\5\S\W\S\2\S\?\S\F\S\+\S\R\S\F', $,
569 IFCLENGTHMEASURE(0.), $);
570 #733= IFCPROPERTYSINGLEVALUE('\S\R\S\Q\S\8\S\=\S\W\S\E\S\5\S\W\S\2\S\?',
571 $, IFCBOOLEAN(.F.), $);

```

(a) File A

```

531 #674= IFCPROPERTYSINGLEVALUE('IsExternal', $, IFCBOOLEAN(.T.), $);
532 #675= IFCPROPERTYSET('3vCh$0ythZzhLClJPLrosJ', #15, 'Pset_WallCommon', $, (
533 #671, #672, #673, #674));
534 #677= IFCRELEDEFINESBYPROPERTIES('260N_OFnjMqInhadQjXR_Y', #15, $, $, (#509),
535 #675);
536 #681= IFCPROPERTYSINGLEVALUE('\S\6\S\(\S\N\S\;\S\O\S\_', $, IFCINTEGER(0),
537 $);
538 #682= IFCPROPERTYSINGLEVALUE('\S\5\S\W\S\2\S\?\S\F\S\+\S\R\S\F', $,
539 IFCLENGTHMEASURE(0.), $);
540 #683= IFCPROPERTYSINGLEVALUE('\S\R\S\Q\S\8\S\=\S\W\S\E\S\5\S\W\S\2\S\?',
541 $, IFCBOOLEAN(.F.), $);

```

(b) File B

Figure 15: The file comparison results using the plain text comparison tool *DiffMerge*.

Table 1: The number of data instances in test cases (M1 – M4).

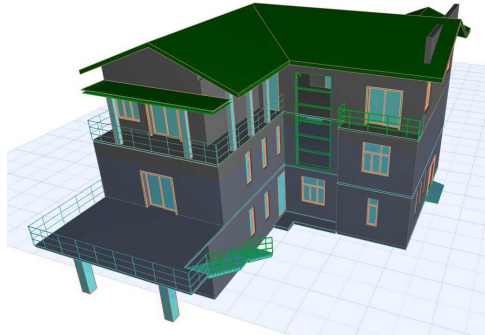
IFC files	#instances
M1 (with redundance)	106,438
M2 (without redundance)	45,461
M3 (with redundance)	103,541
M4 (without redundance)	44,931

548 through removing a window, and it was re-exported as File B (Figure 15(b)).  
 549 Figure 15 shows the corresponding parts of file fragments with the same  
 550 contents but with different instance names. For example, #724 in File A and  
 551 #674 in File B are the identical data instances but with different instance  
 552 names. *DiffMerge* recognizes that the two parts are totally different while  
 553 highlighting their differences. The main reason is that plain text comparison  
 554 methods cannot deal with specific data organization and representation of  
 555 IFC files including the complex referencing and inheritance structures. In  
 556 contrast, our approach recognizes the two parts as the same.

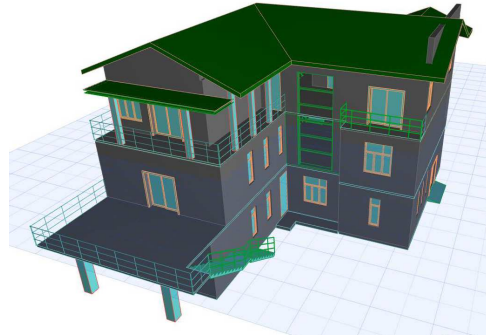
### 557 5.2. Comparison with the flattening-based file comparison method

558 The second experiment compares our method with the flattening-based  
 559 file comparison method [13]. The used IFC files were exported through *Archi-*  
 560 *CAD*, which are referred to as M1 – M4. The corresponding models are  
 561 visualized in Figure 16. In the four test files, M1 contains a large number  
 562 of duplicate data instances, while M2 is the non-redundant file obtained by  
 563 removing the duplicate data instances from M1 (using our *IFCCompressor*  
 564 tool [9]). M3 is obtained by deleting the roof of M1 in *ArchiCAD* and ex-  
 565 porting the file, while M4 is the non-redundant file obtained by removing  
 566 the duplicate data instances from M3. Table 1 shows the number of data  
 567 instances in each IFC file. For instance, the original M1 file contains 106,438  
 568 instances, while the non-redundant M2 file just includes 45,461 instances.

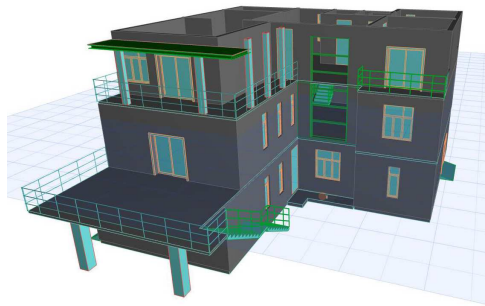
569 Table 2 shows the similarity rates computed using our method and the  
 570 flattening-based method [13]. M1 (with redundance) and M2 (without re-  
 571 dundance) are the same model but with a different number of data instances;  
 572 similarly, M3 (with redundance) and M4 (without redundance) are the same  
 573 model but with a different number of instances. In general, a robust ap-  
 574 proach of IFC comparison should be capable of obtaining a stable similarity  
 575 rate between M3 (or M4) and M1 (or M2). The results in Table 2 suggest



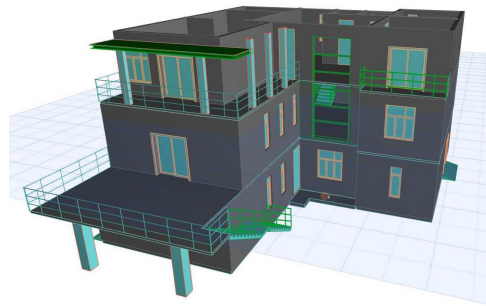
(a) M1 (with redundancy)



(b) M2 (without redundancy)



(c) M3 (with redundancy)



(d) M4 (without redundancy)

Figure 16: Visualizing the models of four test IFC files (M1 – M4). M1 contains a large number of duplicate data instances, while M2 is the non-redundant file through removing the duplicate instances from M1. M3 is the re-exported file after deleting the roof of M1 in *ArchiCAD*, while M4 is the non-redundant file through removing the duplicate instances from M3.

Table 2: The similarity rates computed using our method and the flattening-based method [13].

Target	Source	SR using our method <sup>a</sup>	SR using flattening <sup>b</sup>
M3	M1	83.2%	81.6%
M3	M2	83.2%	81.6%
M4	M1	83.2%	83.2%
M4	M2	83.2%	83.2%

<sup>a</sup>“SR using our method” is the similarity rate computed by our methods.

<sup>b</sup>“SR using flattening” is the similarity rate computed by the flattening-based method [13].

576 that our approach is not sensitive to redundant instances within IFC files,  
 577 which can obtain the consistent similarity rate (83.2%) between M3 (or M4)  
 578 and M1 (or M2). In contrast, the flattening-based method obtains two differ-  
 579 ent similarity rates (81.6% and 83.2%), because of the redundant instances  
 580 within IFC files. As for the flattening-based method, if there are a large  
 581 number of duplicate instances matched in two files, the similarity rate tends  
 582 to be high; otherwise, if there are a large number of duplicate instances not  
 583 matched in two files, the similarity rate tends to be low.

### 584 5.3. Experiments under different parameter conditions

585 The third experiment compares our method under different parameter  
 586 conditions. In Figure 13 and Figure 14, one can select multiple checkboxes  
 587 to ignore the GUIDs, owner history information and the order of property  
 588 set. This can explicitly improve the comparison results. We select two IFC  
 589 files (referred to as M5 and M6) for this test, as visualized in Figure 17. A  
 590 building model was first generated in *ArchiCAD* and then exported as M5.  
 591 Next, M5 was imported into *ArchiCAD* and re-exported as M6 without any  
 592 modification. Before performing file comparison, the duplicate instances of  
 593 M5 and M6 have been removed using our *IFCCompressor* tool [9].

594 Table 3 shows the similarity rates using the *IFCdiff* with different param-  
 595 eters. The similarity rate without any specific parameters is about 85.85%,  
 596 which is the same as the similarity rate in the flattening-based method [13].  
 597 The reason is that the two input files M5 and M6 have no redundant in-  
 598 stances; consequently, the flattening-based method [13] can obtain the same  
 599 result. In Table 3, the similarity rate with ignoring the GUIDs is also 85.85%,

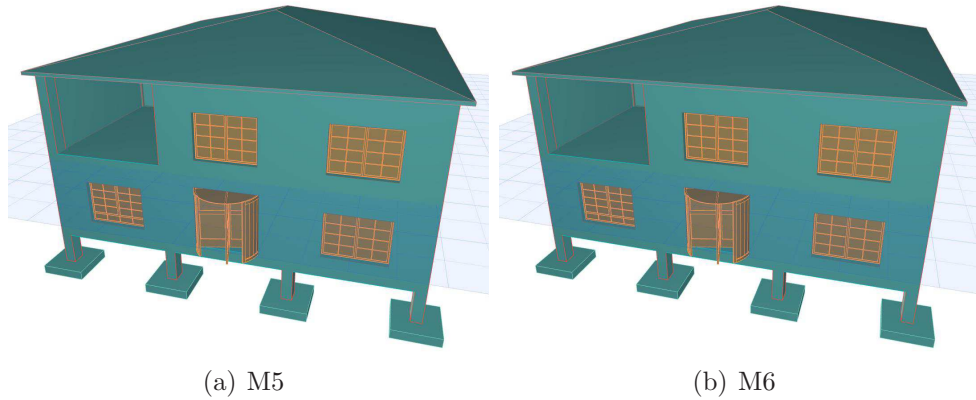


Figure 17: Visualizing the models of M5 and M6. A building model first was generated in *ArchiCAD* and then exported as M5. Next, M5 was imported into *ArchiCAD* and re-exported as M6 without any modification.

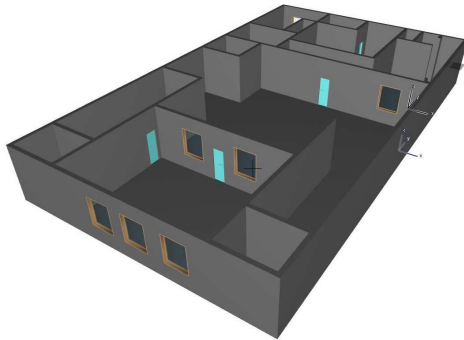
600 which indicates that *ArchiCAD* preserves the GUIDs well during data ex-  
 601 change. Another reason is that the building model was built on *ArchiCAD*  
 602 and was exported as M5 and M6 still through *ArchiCAD*. In other words, the  
 603 GUIDs were generated and maintained by the same system (i.e. *ArchiCAD*)  
 604 itself. However, the GUID preservation rate often is low when a model is  
 605 imported and exported in two different systems, as illustrated by Lee et al.  
 606 [13].

607 In this table, the similarity rate with ignoring owner history information  
 608 is about 95.12%, which is highest in this table. The reason is that a large  
 609 number of data instances cite the entity *IfcOwnerHistory* which holds the  
 610 modeler and modeling software information. The owner history information  
 611 changes whenever a file is imported and exported from a system, even if no  
 612 revisions are made to the model. Therefore, when ignoring the changes of  
 613 owner history information, the similarity rate can be improved significantly.

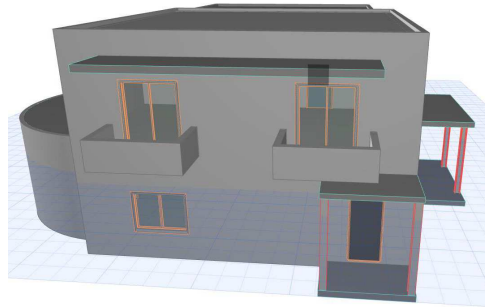
614 Finally, we test the similarity rate while ignoring the order of properties  
 615 in property sets, which is about 86.26% better than the default (i.e. 85.85%).  
 616 This suggests that the orders of some attribute instances have been changed  
 617 during the data exchange process, even if importing and exporting was done  
 618 in the same system (i.e. *ArchiCAD*). The reason is the difference of model  
 619 mapping mechanism between the internal models of BIM software platforms  
 620 and the standard IFC data model, where the properties of some objects held  
 621 in *ArchiCAD* are in different order to that in the IFC data model.

Table 3: The similarity rates using the *IFCdiff* with different parameter conditions.

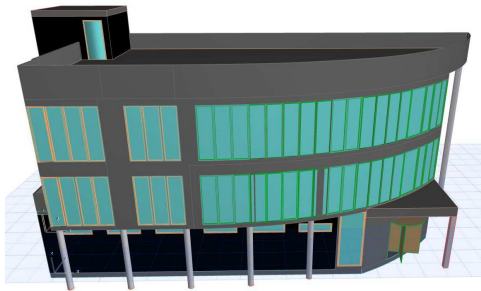
Parameters	Similarity rate
Default (without any specific parameters)	85.85%
Ignore the GUIDs	85.85%
Ignore owner history information	95.12%
Ignore the order of property set	86.26%



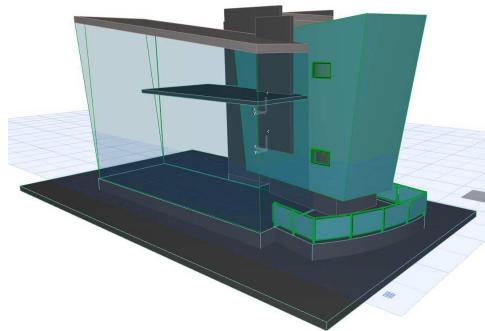
(a) M7



(b) M8



(c) M9



(d) M10

Figure 18: Visualizing the four models (M7 – M10) used for testing computational time and space.



622 5.4. Computational time and space

623 5.4.1. Computational time

624 The fourth experiment compares computational time and space between  
 625 our method and other methods [13]. In this test, four building models were  
 626 developed in *Autodesk Revit 2014* and exported as the initial IFC files (re-  
 627 ferred to as M7 – M10), as visualized in Figure 18. Then we import the four  
 628 files into *Revit* and *ArchiCAD*, and export them as new IFC files without  
 629 making any changes to the models. The new IFC files are renamed M7\_R,  
 630 M8\_A, M9\_R and M10\_A, where “\_R” and “\_A” denote that the files are  
 631 exported through *Revit* and *ArchiCAD*, respectively. The new files are used  
 632 as the target files, while the initial files are used as the source files.

Table 4: The details of paired IFC files used for testing computational time and space.

No.	Target files			Source files		
	Name	Size(MB)	#instances	Name	Size(MB)	#instances
1	M7_R	0.591	11,753	M7	0.425	8,287
2	M8_A	1.040	24,277	M8	1.257	26,234
3	M9_R	2.519	42,884	M9	3.511	68,114
4	M10_A	9.817	215,354	M10	4.364	91,023

633 Table 4 gives the details of those files to be compared, where “No.” is  
 634 the index of paired files to be compared, “Size(MB)” is the file sizes, and  
 635 “#instances” is the number of data instances within the files. Note that  
 636 although there is no modification in the imported and exported models, the  
 637 file sizes and the number of data instances still suffer some changes. For  
 638 example, M10 contains 91,023 instances, while there is a great increase of  
 639 instances in the re-exported M10\_A (215,354 instances). The main reason is  
 640 that different systems map data into the IFC files in different ways.

641 Next, comparisons of paired files are made using the flattening-based  
 642 method [13] and our approach, respectively. Table 5 shows the computa-  
 643 tional time of the two methods. In Ref. [13], the process of file comparison mainly  
 644 contains three steps: (1) parsing data instances into the memory, (2) flatten-  
 645 ing all the instances and (3) comparing pairs of instances. In Table 5, we list  
 646 the computational time of each step of the flattening-based method and the

Table 5: Computational time of two methods for the paired IFC files in Table 4.

No.	Flattening-based method				Ours	$RT^a(\%)$
	Parse(s)	Flatten(s)	Compare(s)	$T_1^b$ (s)	$T_2^c$ (s)	
1	0.3276	2.0592	7.7377	10.1245	1.3728	86.44%
2	0.3276	2.0436	43.8987	46.8159	4.8048	89.74%
3	2.7924	4.6956	254.5	261.988	8.5333	96.74%
4	7.1605	135.861	1702.0	1845.02	81.2609	95.60%

<sup>a</sup>“ $RT$ ” is the percentage of reduced time when using our approach.

<sup>b</sup>“ $T_1$ ” is the total time of the flattening-based method.

<sup>c</sup>“ $T_2$ ” is the time of our approach.

647 total time (“ $T_1$ ”). In addition, the time of our approach is given by “ $T_2$ ”, and  
 648 the percentage of reduced time is listed by “ $RT$ ”, where  $RT = (T_1 - T_2)/T_1$ .  
 649 The result in Table 5 shows that our approach can significantly reduce the  
 650 time in the file comparison process. For example, the percentage of reduced  
 651 time is about 95.60% for the comparison of M10\_A and M10. As mentioned  
 652 in Section 2.5.3, the flattening-based method is often time-consuming for  
 653 comparison of large IFC files, especially with numerous duplicate instances.  
 654 It takes a lot of time to perform the two steps of flattening and compar-  
 655 ing in [13]. In contrast with the flattening-based method, our approach has  
 656 an advantage when dealing with large file comparison. In this experiment,  
 657 for example, the flattening-based method costs 1845.02s for the comparison  
 658 of M10\_A and M10, while our approach just takes 81.2609s to process the  
 659 comparison of the same files (reducing 95.60%). The result shows that the  
 660 percentage of reduced time with our algorithm is generally very high (the  
 661 average is 92.13%) for tested cases.

#### 662 5.4.2. Computational space

663 In general, the flattening process in [13] also increases the size of an IFC  
 664 file several times or even dozens of times. Table 6 shows the size of original  
 665 target files in Table 4 and the size variation after running the flattening-based  
 666 method and our approach. As seen in this table, when using the flattening-  
 667 based approach, the sizes of some files are increased by more than ten times.  
 668 In contrast, our approach reduces the file to a smaller size, which greatly  
 669 improves the comparison efficiency.

Table 6: Space requirements of two methods in the file comparison process.

No.	Original <sup>a</sup> (MB)	Flattening-based method		Ours	
		Flatten <sup>b</sup> (MB)	Increase <sup>c</sup> (%)	Process <sup>d</sup> (MB)	Reduce <sup>e</sup> (%)
1	0.591	33.292	5533.16%	0.466	26.82%
2	1.040	9.407	804.52%	0.722	44.04%
3	2.519	31.959	1168.72%	1.905	32.23%
4	9.817	70.175	614.83%	6.398	53.44%

<sup>a</sup>“Original” is the size of original target files in Table 4.

<sup>b</sup>“Flatten” is the size of space after flattening all instances in the target files.

<sup>c</sup>“Increase” is the percentage of increased space when using the flattening-based method.

<sup>d</sup>“Process” is the size of space after removing redundant instances using our approach.

<sup>e</sup>“Reduce” is the percentage of reduced space when using our approach.

670 In addition, the flattening-based method needs to use all the instances  
671 for comparison. Unlike that, since enormous duplicate instances are removed  
672 from the original files based on our approach, the number of actual instances  
673 used for comparison is significantly decreased. Table 7 lists the number of  
674 instances used for comparison based on the flattening-based method and our  
675 approach. The result shows that the average percentage of reduced instances  
676 using our approach reaches 25% for the tested cases.

Table 7: Counting the number of instances used for comparison based on the flattening-based method and our approach.

No.	Target files	$N_1^a$	$N_2^b$	Reduce <sup>c</sup> (%)
1	M7_R	11,753	10,005	14.87%
2	M8_A	24,277	16,393	32.48%
3	M9_R	42,884	31,911	25.59%
4	M10_A	215,343	154,617	28.20%

<sup>a</sup>“ $N_1$ ” is the number of instances used for comparison based on the flattening-based method.

<sup>b</sup>“ $N_2$ ” is the number of instances used for comparison based on our approach.

<sup>c</sup>“Reduce” is the percentage of reduced instances when using our approach.



Figure 19: Visualizing the IFC model used in a real-life case.

677 *5.5. Preliminary test in a real-life case*

678 In order to test the performance of our approach in a real-life case, an  
679 apartment building model in the Yunnan province in China is selected as a  
680 preliminary test. The architectural design model was developed in *Revit*, and  
681 exported as an IFC file. This selected model has been used for our previous  
682 case studies including IFC-based path planning [8] and IFC compression [9].  
683 The original IFC file is about 156.0 MB, which includes more than 2.8 million  
684 data instances with numerous duplicate instances. The corresponding model  
685 is visualized in Figure 19.

686 In this case study, we first remove all duplicate instances from the origi-  
687 nal file using the *IFCCompressor* tool [9]. Then the newly non-redundant  
688 file is compared with the original file, which produces the similarity rate of  
689 100%. The result suggests that our approach is not sensitive to redundant  
690 instances even in large IFC files. When using our *IFCdiff* tool, the time  
691 cost of comparison process is about 371.1s. In contrast, the flattening-based  
692 method fails to handle such large IFC files.

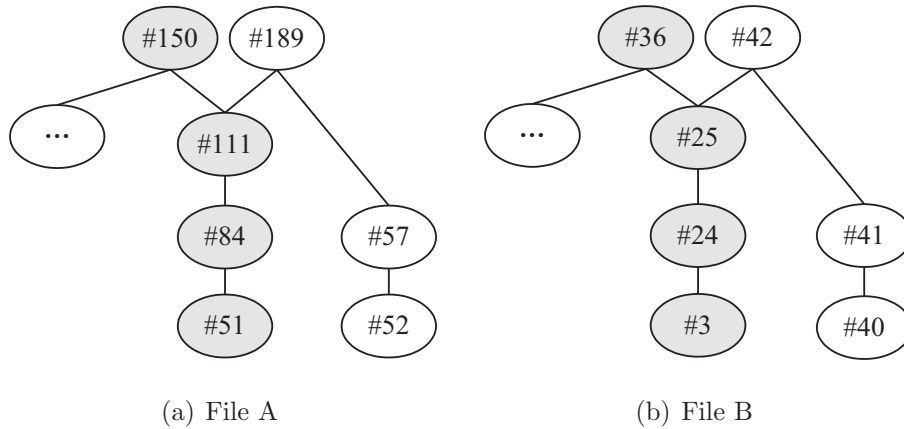


Figure 20: An example for illustrating the incremental backup content, where File A is the previous version and File B is the current version. The identical data instances are highlighted by the light gray nodes, while the different data instances are the white nodes.

693 5.6. Application to incremental backup of IFC files

694 One potential application of our approach is for incremental backup of  
 695 IFC files. An *incremental backup* is a type of data backup that backs up  
 696 only the new or changed data since the last incremental backup. The design  
 697 and management of building models follow an iterative process, which often  
 698 includes frequent revision and updating on one or more basic models during  
 699 the lifecycle of a construction project. This requires an effective method for  
 700 IFC file backup. Time and disk space can be saved by only backing up the  
 701 changed data.

702 The traditional *full backup* backs up all data on a disk even if minor  
 703 changes are made to the files, which is time-consuming and space-intensive  
 704 for IFC data management. Therefore, incremental backups are often desir-  
 705 able as they consume smaller storage space and are quicker to perform than  
 706 full backups. Although pure text comparison methods can be directly used  
 707 for incremental backup of IFC files, they cannot deal with specific data or-  
 708 ganization and representation of IFC files, as mentioned in Section 5.1. As  
 709 a result, numerous consistent data instances (with different instance names)  
 710 are considered to be different, and the size of incremental backup data is  
 711 often close to the full backup.

712 The incremental backup mainly consists of identifying and recording the  
 713 changed data since the last backup. Our comparison approach can be directly

```

...
#40 = IFCCARTESIANPOINT((0., 0., -150.));
...
#41 = IFCAXIS2PLACEMENT3D(#40, $, $);
...
#42 = IFCLOCALPLACEMENT(#38, #41);
...

```

Figure 21: Illustrating the differences between the previous and current versions, which are saved to record the changed data in incremental backup.

Table 8: Comparison of storage space between the full backup and our incremental backup.

Previous	Current	Full backup(MB)	Incremental backup(MB)	Reduce(%)
M2	M4	2.231	0.603	72.97%

714 applied to identify the differences between two IFC files. Then the differences  
715 are saved as the portion that has changed. Figure 20 shows an example for  
716 illustrating the incremental backup content, where File A is the previous  
717 version and File B is the current version. Here the identical data instances  
718 are highlighted by the light gray nodes, while the different data instances are  
719 the white nodes. Finally, the differences between two files are saved to record  
720 the changed data since the last incremental backup, where we typically save  
721 the different data instances and the hash table of matching instances in a  
722 specific file form (see Figure 21).

723 We also test our incremental backup strategy on two actual files (M2 and  
724 M4), as shown in Figure 16. Table 8 shows the comparison of storage space  
725 between the full backup and our incremental backup, where M2 and M4 are  
726 assumed to be the previous and current versions, respectively. The result  
727 suggests that our incremental backup saves around 73% space in contrast to  
728 the full backup. The incremental backup of IFC files is an attractive research  
729 topic, and its full implementation including an efficient recovery process will  
730 be left to our future work.

## 731 6. Conclusion, contribution and discussion

732 This paper presents a content-based automatic comparison approach for  
733 IFC files, and presents the development of a file comparison tool *IFCdiff*.  
734 The novelty is to build the hierarchical structures for comparing IFC files  
735 along with eliminating their redundant instances in the comparison process.  
736 Here the built hierarchical structures of two IFC files are compared with an  
737 iterative bottom-up procedure instead of comparing the original files. Such  
738 a process does not need to flatten all the data instances in IFC files. In the  
739 comparison process, all matching instances between two files are saved in the  
740 hash table, while the differences between them are also recorded for further  
741 applications. To evaluate the performance of our approach, the presented  
742 approach is tested on some IFC files exported through several commercial  
743 BIM software platforms. Finally, we make use of the presented approach  
744 to demonstrate a potential application to incremental backup of IFC files.  
745 The experimental results show that our approach outperforms the previous  
746 methods.

747 The significant contributions of our work are summarized as follows.

- 748 - We build the hierarchical structures for comparing IFC files with an iter-  
749 ative bottom-up procedure. Compared with the previous flattening-  
750 based approach, our approach avoids the procedure of flattening in-  
751 stances. As a result, our approach can greatly reduce the computa-  
752 tional time and space in the file comparison process. The experimental  
753 result shows that the percentage of reduced time with our algorithm is  
754 generally very high (the average is 92.13%) for tested cases.
- 755 - In the level-by-level comparison of hierarchical structures, we also e-  
756 liminate redundant instances appearing in two IFC files. This brings  
757 two advantages. On the one hand, by removing redundant instances  
758 can significantly decrease the number of data instances to be com-  
759 pared, which improves the comparison efficiency. On the other hand,  
760 by removing the redundant instances while keeping the complete IFC  
761 models, the comparison result using our approach is not sensitive to  
762 redundant instances in IFC files, which brings a stable and reliable  
763 similarity superior to the previous methods.
- 764 - We apply the presented comparison approach to incremental backup  
765 of IFC files. Here our approach is used for identifying and recording  
766 the changed data between the previous version and the current one.

767       The result suggests that our incremental backup can greatly save the  
768       storage space in contrast to the full backup.

769       Some previous studies have contributed to the issue of removing redun-  
770       dant instances in a single IFC file, such as Solibri IFC Optimizer [27] and  
771       IFCCompressor [9]. In this paper, we follow a similar manner to [9] to remove  
772       redundant instances in each level comparison, so the strategy of removing  
773       redundant instances used in this paper is not new. In practice, however,  
774       the IFC files generated by various software platforms often include a large  
775       number of redundant instances [9, 13], so the redundant instances should  
776       be considered in the process of IFC comparison. In this paper, we argue  
777       that it is meaningful and important to make use of the hierarchical struc-  
778       tures for comparing IFC files along with removing redundant instances. This  
779       combination of hierarchical comparison and redundancy elimination can sig-  
780       nificantly speed up the file comparison process and obtain a stable similarity.  
781       Even if two IFC files without redundancies are compared to each other, our  
782       approach can still reduce the computational time and space in contrast with  
783       the previous flattening-based approach. The *IFCdiff* presented in this paper  
784       can be considered as a complementary tool for the existing IFC tools.

785       Our comparison method only deals with the *syntax* content of data in-  
786       stances explicitly extracted from the input IFC file itself, but the *semantic*  
787       content of data instances implicitly derived from the IFC file is not handled  
788       yet. The domain of geometry comparison and objectified relationships be-  
789       longs to the semantic comparison problem, which is quite complicated and  
790       will be our future work. We give an example for illustrating this problem as  
791       follows. In our method, one data instance explicitly extracted from the input  
792       IFC file itself consists of three terms (i.e. instance name, entity name and  
793       attribute values). If the entity name and attribute values between two data  
794       instances are consistent, they are considered to be the same. In contrast,  
795       if the geometric representation is different between two data instances, they  
796       will be potentially considered to be different in our method. This may not  
797       be always true. In particular, the IFC schema provides various geometric  
798       representations (such as swept, CSG and B-rep) for a solid model, which can  
799       be freely chose by an BIM modelling system. This means that the same solid  
800       model may have many different geometric representations. To address this  
801       issue, a possible way of geometry comparison is to first discretize the solid  
802       models into 3D meshes and then use some existing 3D shape comparison  
803       methods for the discretized shapes [28, 29, 30, 31, 32, 33].



## 804 **Supplementary material**

805 The online *IFCdiff* tool and its demonstration can be accessed at: [http:](http://cgcad.thss.tsinghua.edu.cn/liuyushen/ifcdiff/)  
806 [//cgcad.thss.tsinghua.edu.cn/liuyushen/ifcdiff/](http://cgcad.thss.tsinghua.edu.cn/liuyushen/ifcdiff/).

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## 811 **Appendix A. Comparison of computational complexities**

812 In this section, we mainly compare computational complexities between  
813 our approach and the flattening-based algorithm [13]. If only considering  
814 the comparison process, both the time complexities of our method and the  
815 flattening-based method are  $O(n \log(n))$  (see Section 4.4) with the assump-  
816 tion that the length of each data instance is a constant, where  $n$  is the average  
817 number of data instances between two IFC files to be compared. Our method  
818 is able to accomplish the comparison of reference instances by only compar-  
819 ing the instance name with the help of an auxiliary map storing matched  
820 instance names. However, the flattening-based method replaces all the ref-  
821 erence numbers with their actual values in each IFC file, which makes an  
822 IFC file into a structure that does not include any referencing or inheritance  
823 structure. This will result in that the length of the flattened data instance  
824 increases dramatically (see an example in Section 2.4). Therefore, the length  
825 of the data instance cannot be seen as a constant any more. Assuming that  
826 the average length of the flattened data instance is  $L$ , the time complexity  
827 of the flattening-based algorithm is  $O(Ln \log(n))$ , which of course will take  
828 much more time than our method.

829 In addition, our algorithm has some extra advantages in contrast to the  
830 flattening-based method from the perspective of space complexity. In order to  
831 facilitate the description and analysis, the comparing file and the compared  
832 file can be simplified as a hierarchical structure to establish the reference  
833 mechanism based on a tree. Let  $d$  be the depth of the tree and  $L$  be the  
834 average length of each data instance. We also assume that the number of  
835 attributes of the instance is  $K$  and they are all references. Finally, from the  
836 bottom of the tree, each data instance is referenced by others for  $h$  times.

837 In our algorithm, the number of the terminal level is  $k^{d-1}$  and the total  
838 length is  $k^{d-1}L$ . Similarly, the second layer from the bottom has  $k^{d-2}$  data  
839 instances and the total length is  $k^{d-2}L$ . Therefore, we can deduce that the  
840 space complexity of our algorithm is

$$S_1 = k^{d-1}L + k^{d-2}L + \dots + L. \quad (\text{A.1})$$

841 Meanwhile, in the flattening-based method, the number of the terminal  
842 level is  $k^{d-1}$  and the total length is  $k^{d-1}L$ . Similarly, the second layer from the  
843 bottom has  $k^{d-2}$  data instances and the total length is  $k^{d-1}Lh + k^{d-2}L$ . Thus,  
844 we can deduce that the space complexity of the flattening-based algorithm is

$$\begin{aligned} S_2 &= k^{d-1}L + (k^{d-1}Lh + k^{d-2}L) + (k^{d-1}Lh + k^{d-2}L)hL + k^{d-3} + \dots + L \\ &= k^{d-1}Lh + (k^{d-1}Lh + k^{d-2}L) + \dots + k^{d-1}L + k^{d-2}L + \dots + L \\ &= k^{d-1}Lh + (k^{d-1}Lh + k^{d-2}L) + \dots + S_1. \end{aligned} \quad (\text{A.2})$$

845 Apparently, the whole file increases dramatically large after the flattening  
846 process, which explains why the flattening-based method takes much longer  
847 than our method.

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