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IMISketch: an interactive method for sketch recognition

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8 Abstract

1

2

In this paper, we present a new generic method for an interactive interpreta-9 tion of sketches to avoid a fastidious verification phase. After a preprocess-10 ing phase in which we extract a set of primitives, the interpretation process 11 consists of an interactive analysis. The analyzer is based on a competitive 12 breadth-first exploration of the analysis tree. As opposed to well known 13 structural approaches, this method allows to evaluate simultaneously several 14 possible hypotheses of recognition in a dynamic local context of document. 15 The decision process is able to solicit the user in the case of strong ambiguity: 16 when it is not sure to make the right decision. The user explicitly validates 17 the right decision. While, in practice these approaches often induce a large 18 combinatory, this paper presents optimization strategies to reduce the com-19 binatory. The goal of these optimizations is to have time analysis compatible 20 with user expectations. These strategies have been integrated into both pre-21 processing and analysis phases. To validate this interactive analysis method, 22 several experiments are reported in this paper on off-line handwritten 2D 23 architectural floor plans. 24

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²⁵ Keywords: sketch recognition, breadth-first exploration, interactive

²⁶ recognition, 2D architectural floor plans, Kalman filter

27 1. Introduction

Nowadays, digital documents are becoming more and more omnipresent in 28 our life. Many reasons, such as the flexibility provided by digital processing, 29 have led to transform handwritten documents to digital ones. In this context, 30 people are working on mapping technical paper documents, like architectural 31 floor plans, to digital ones. We aim at offering a complete and homogeneous 32 solution to unify paper document recognition and pen-based composition (for 33 instance: with Tablet PC). We present the IMISktech system : Interactive 34 Method for Interpretation of Sketches. The input of this system is a scanned 35 image of handwritten architectural plan and after interpretation the output 36 is its digital version. This method is the result of four years of research 37 that leads to several scientific publication [14, 16, 17, 15]. This paper is a 38 synthesis of this method. We focus on the extraction of primitives that feed 39 the analyzer to optimize the management of combinatory. 40

We have identified two major approaches for document analysis: syntactic and statistical approaches. Choosing one of these two approaches depends on the document type.

The syntactic approaches [6, 8, 29, 12, 21, 22] lean on prior knowledge of the document structure to drive the analysis. They are often based on visual languages for describing this knowledge and generating the analyzer. However, syntactic methods have difficulties to incorporate the uncertainty. The statistical approaches [25, 30, 32] provide a better ability to incor⁴⁹ porate uncertainty and usually lack the ability to convey the hierarchical ⁵⁰ structure of the document. The use of statistical approaches needs a wide ⁵¹ learning on a homogeneous and labeled base. Each type of approach has ⁵² advantages and drawbacks. The interpretation of handwritten structured ⁵³ documents needs on the one hand an approach that retains its structure, ie ⁵⁴ a syntactic approach, and on the other hand an approach that provides a ⁵⁵ better ability to incorporate uncertainty, ie a statistical approach.

In this work, we design a complete system for the sketch interpretation: 56 IMISketch¹. One of the main originalities of IMISketch is to avoid an a pos-57 *teriori* verification phase by soliciting the user. After a preprocessing phase 58 in which primitives of the structured document are extracted, the system 59 is characterized by an interactive analysis phase. The analyzer (referred as 60 IMISketch) uses a new syntactic approach based on an interactive and lazy 61 interpretation of the document. Unlike the classical syntactic approaches, 62 IMISketch does not always select the first or the best found hypothesis. The 63 associated analysis process is able to take into account the uncertainty. 64

Thanks to the interactivity, the user can be solicited, if needed, by the analyzer to raise ambiguities of recognition [17] i.e. to choose between two or more possible hypotheses or to enrich the *a priori* knowledge of the system [14]. In fact, the user participation has a great impact to avoid error accumulation during the analysis step. To detect ambiguities, we should adopt a method based on breadth-first exploration. Like all the analysis methods based in breadth-first exploration, this approach can induce a large

¹Interactive Method for Interpretation of Sketch

⁷² combinatorics. This combinatorics mainly depends on the quality of prim-⁷³ itives extracted from the image and the manner to analyze them. In this ⁷⁴ paper we propose some optimizations to reduce it, by addressing these two ⁷⁵ points. These optimizations are introduced from the phase of segmentation ⁷⁶ to the analysis. They will lead to the new system IMISketch+.

The complete system can be applied to off-line documents (image), as illustrated in this paper (Figure 9(a)), as well as on-line or vectored documents.

In the state of the art, one interesting generic approach is the LAD-80 DER [19] [18] system which has been proposed by Hammond and Davis 81 for interpreting a *posteriori* or on the fly on-line handwritten documents. 82 LADDER language has been exploited for the design of various systems of 83 interpretation of structured documents, such as UML [18], electrical dia-84 grams [2] or complex graphs [20]. Also Plimmer proposed InkKit that is a 85 framework and a toolkit to recognize complex components [31] [13]. In addi-86 tion VR Sketchpad [27] is a pen-based computing environment for inputting 87 and locating 3D objects in a virtual world. 88

Unlike these methods, our method interprets off-line handwritten struc-89 tured documents. It has been tested on 2D architectural floor plans. The 90 specific task of floor plan analysis has been addressed for more than twenty 91 years. Lladós [26] proposed a method for understanding hand drawn floor 92 plans using subgraph isomorphism and Hough transform. Aoki [3] proposed 93 also a method for interpreting a hand-sketched floor plan. This method fo-94 cuses on understanding the hand sketched floor plan and converting it into 95 a CAD representation. Also, Ahmed [33] proposed an analysis method spec-96

⁹⁷ ified in printed architectural floor plans.

⁹⁸ Contrary to these methods that can require a fastidious *a posteriori* veri-⁹⁹ fication phase, IMISketch system attempts to avoid this phase by integrating ¹⁰⁰ the user during the analysis process.

The recognition of a structured document using a structural approach 101 needs an a priori description. The modeling of structured documents differs 102 from one type to another. Several techniques allow the document descrip-103 tion. Yamamoto [35] and Bunke [5] use classical one-dimensional grammars. 104 Other techniques are used to model two-dimensional documents. Fahmy [11] 105 and Bunke [4] offer grammar graphs. These grammars have been widely used 106 in the various communities for interpreting off-line documents such as math-107 ematical formulas. Despite graph grammars offer a very expressive mecha-108 nism for pattern recognition, these grammars have their limitations. They 109 are expensive to implement and difficult to handle by the developer, espe-110 cially when the productions become numerous. These graphs are also poorly 111 adapted to deal with uncertainty. 112

Our goal is to analyze documents of different kinds such as handwritten 113 documents. To overcome this problem, we adopt context-driven constraint 114 multi-set grammars (CD-CMG), designed for on-line recognition [28] asso-115 ciated with a scoring approach based on the fuzzy logic theory. The main 116 contribution of our work is to modify and to extend this formalism to design 117 an interactive analyser for off-line recognition. This strategy allows to solicit 118 the user in the case of strong ambiguity and avoids the fastidious a posteriori 119 verification task to find and correct the remaining interpretation errors. 120

The remaining of the paper is organized as follows. In the section 3(a),

we introduce the architecture and the basic principles of IMISketch method. The phase of primitive extraction is described in section 3. Section 4 presents the concepts that are linked with an interactive breadth-first analysis. In section 5 implementation and optimization of IMISketch analyzer are presented. Experimental results on interpretation of images of 2D handwritten architectural floor plans are reported in section 6 and finally, section 7 concludes the paper.

129 2. Interactive analysis stages

In this section, we summarize the different steps of treatment to ensure the 130 recognition of a handwritten structured document (cf. Figure 1). The first 131 step is the segmentation process. This step is purely off-line (i.e. without user 132 interaction). The aim of this phase is to extract all the basic primitives that 133 will be used to analyze the document. In the context of sketch recognition, 134 the segmentation process consists in extracting handwritten strokes as a set 135 of segments. This part is detailed in section 3. The second main step is to 136 analyze these primitives and to compose them according to their structural 137 arrangement in the document to identify the symbols. Our analyzer is made 138 of two associated key processes: the recognition of the document structure 139 is managed by **the grammatical analyzer** that drives the calls of **symbol** 140 **classifiers** to evaluate a fuzzy scoring for each hypotheses. For instance, in 141 architectural plans, structural recognition detects walls, opening, etc., and 142 the fuzzy classifiers identify the opening to window, door, etc. 143

The user can be solicited during the analysis process in case of ambiguity detection. This approach is generic and needs some *a priori* knowledge at the structural level by defining a grammar that will describe the document
structure and at the classifier level to identify the symbols. This knowledge
corresponds to the specific part of the analysis system.

We detail in the following sections the preprocessing (section 3) and the analysis phases (section 4).



Figure 1: Synthetic scheme for structured documents interpretation

¹⁵¹ 3. Preprocessing phase: primitive extraction

152 3.1. Related works

The recognition of architectural plans has already been studied in particular by Dosch et al [10]. In this kind of plans, segments, representing the walls, are primitives widely used. In these works, most of the analyzed plans have been drawn with a ruler (or CAD software), consequently, segments are really straight. In [23], Hilaire proposed a method that improves the method of Dosch. The originality of his work comes from the segmentation process of the skeleton. This aspect is very important in the context in which drawings are crossing. His paper contains an important bibliography. Hilaire, as Dosch, works on printed documents, but our aim is to process handwritten documents. In such documents, walls are not straight lines and the representation of doors and windows use arcs of a circle which are sometimes very roughly drawn.

In [7], Chang proposes to carry out the vectorization of handwritten documents by Bezier curves. This approach has been designed for the vectorization of cartoons, but it does not lead to an easy representation to be used for a later interpretation of symbols like pieces, doors, windows, etc.

In [9] de Brucq uses a Kalman filter to decompose handwritten strokes into straight lines and arcs of circle. These primitives are well adapted for architectural plan interpretations, but the system works with an on-line representation of the drawing.

173 3.2. Our method

To extract lines in off-line sketch documents, we propose to adapt the method we presented in [24]. It uses a Kalman filter.

A Kalman filter is a prediction verification process that provides an estimate of a model from observations. In the case of the drawing extraction, our model is based on three values: the thickness, the position and the local slope of the line. When the drawing is relatively regular, the Kalman filter can use these three values to predict the next position and update the model from the pixels of the image. In areas that include intersections, the observed thickness is not consistent with the predicted thickness, but it is possible to use the prediction ability of the filter to pass through these regions. In a region where the direction of the line changes rapidly, it is not possible to find black pixels in the predicted position and the follow-up of the drawing stops. Equations and more details can be found in [24].

For sketch documents the most important parameter for the Kalman filter 187 is the one that defines the variability of the slope. With a large value (0.1, i.e.188 a 10% variation of the slope) a curved section of drawing is detected as a single 189 segment joining its two ends (Figure 2(a)). With a very low value (0.005) 190 the model allows a very little variation of the slope, and thus many small 191 segments are detected in the same curved section (Figure 2(b)). We note, in 192 this example, that the symbols representing opening (door or window) are 193 more precisely represented in Figure 2(b) (precise decomposition) than in 194 Figure 2(a) (rough decomposition). A precise decomposition increases the 195 number of segments extracted from the image, however this decomposition 196 can be very useful when using a classifier to recognize symbols. 197

To fix ideas, the full plan from which is extracted the Figure 2 is represented by 144 segments for the rough decomposition and 177 segments for the precise one. This increase of the number of segments is mostly due to small segments in the opening regions. We must notice that a little region with many small segments will produce a combinatorics explosion of the analyze time. This effect is pointed out later in the section 6 and Figure 9(c).

To overcome this dilemma, i.e. a precise representation but a low combinatorics, we propose to extract only a precise decomposition and to build spatial relationships between the segments. Thus, it is important to remember if a segment is a curvilinear extension of another one. During the



Figure 2: Extraction of primitives: Original drawing in light gray, primitives shown in dark black

tween consecutive segments)

detection of segments by the Kalman filter, it is possible to detect that a 208 part of the drawing belongs to two different segments (Figure 2(c)). In this 209 case, we remove this common part from one of the two segments and mem-210 orize that their is a curvilinear link between the extremities of these two 211 segments. The small circles in Figure 2(d) represent a connection between 212 two segments. Each set of segments attached to one another corresponds to 213 a polygonal approximation of the curved parts of the drawing (for instance 214 an opening). 215

Thus, there are two types of primitives extracted from an image: sim-216 ple segments and polygons. We transmit the connection information so that 217 the interpretation process can directly use simple segments and polygons as 218 primitive, or can decide to split a polygon into segments according to some 219 structural information like the size of the polygon or criterion of collinearity. 220 The use of this dual representation on one hand allows a more precise char-221 acterization of curve sections and on the other hand limits the combinatorics 222 during the interpretation process. 223

224 4. Interactive breadth-first exploration

In this section, we present the analyzer by first describing its main characteristics. Then, we detail the different steps of the internal analysis. The description of this method is followed by a concrete example in the next section (section 5).

229 4.1. Analyzer characteristics

The interpretation strategy of structured documents is driven by the *a priori* knowledge we have on the structural rules of the application domain. These structural information are formalised by a visual language from which we can automatically develop a breadth-first interactive analyser. For handwritten architectural plans, the interpretation of a primitive takes into account the neighbouring objects. Moreover, as we work in a two-dimensional context (images), it can induce a large combinatorics. To overcome this complexity problem, we propose to guide our breadth-first exploration using a spatial contextual focus.

This idea is close to the well known LL(k) analysis, where reading the next 239 k tokens enables to choose without ambiguities which rule must be applied. 240 In the same manner, in our two-dimensional analysis, we have to limit the 241 number of token k to explore, i.e. the depth of the analysis. Consequently, 242 contrary to the LL(k) analysis, the exploration of the following tokens does 243 not allow to take a unique decision on the rule to apply, because we volun-244 tarily limit the value of k. Moreover, sometimes, the grammar is not LL(k) 245 for any k and the analyser meets ambiguities. In those two cases, the process 246 can not be sure to take itself the decision, and may hesitate between several 247 hypotheses. In order to validate the right decision, we propose an analysis 248 process that can, through its decision process, take the right decision or so-249 licit the user in case of ambiguities. This analyzer is based on the following 250 characteristics: 251

- 252 253
- the expression of a priori structural knowledge of the document through a visual language based on production rules;

• a two-dimensional descending breadth first analysis;

255

• a spatial contextual focus of the exploration to limit the combinatorics;

- the formalization of uncertainty by the attribution of scores to each 256 hypothesis, represented by a tree analysis branch;
- some user solicitations: if the ambiguities can not be resolved in the 258 local context in an automatic manner, the user will be solicited by the 259 analyzer to resolve the ambiguity. 260

These characteristics have been defined to ensure a good interaction between 261 the process of analysis and the user. This interactivity allows in particular 262 to avoid an *a posteriori* verification phase, which can become fastidious on 263 complex documents. Indeed, the user participation, on the critical phases of 264 the analysis of the document, has a great impact to avoid error accumulation 265 during the analysis step and overcomes the combinatorics due to the sketch 266 complexity. 267

4.2. Steps of the analysis 268

257

The analyzer tries to match the set of primitives contained in the doc-269 ument with the structure model defined by the two-dimensional production 270 rules. The interactive breadth-first analysis process consists of three stages: 271 1) defining the local context, 2) building the analysis tree and 3) making the 272 decision. The inner part(*) of Figure 1 illustrates these three phases and the 273 relationships between them. 274

4.2.1. Defining the local context 275

The recognition of a given primitive depends on its neighbourhood in 276 structured documents. The analyzer begins by defining a spatial contextual 277

focus that aims to limit the combinatorial exploration due to the breadth-278 first exploration of tree analysis. The structured document requires a two-279 dimensional context. This two-dimensional local context is defined for an 280 analysis tree as the maximum distance between the elements of the root and 281 the elements of any leaves. The choice of the size of the local context depends 282 on the application domain. For example, to interpret an architectural plan, 283 we suggest a local context with a size corresponding to the maximum size of 284 an opening (door, window, etc.). Figures 3(a) and 3(b) show the shifting of 285 the local context in two consecutive steps. Once the local context is set, we 286 go to the building of the analysis tree stage. 287





(b) Local context at step s+1

Figure 3: Dynamic adaptation of local context around the element to analyse

288 4.2.2. Building the analysis trees

In this stage, the analyzer explores all the possible hypotheses of inter-289 pretation in the spatial context using a set of two-dimensional rules that 290 describe the structure of the document. Each primitive can be interpreted 291 in several ways. Each node or leaf is the application of a production rule 292 deduced from the previous node. Every leaf or node of the tree has a score 293 calculated from both its local score and the score obtained from the preceding 294 nodes. Every score determines the adequacy degree to validate a production. 295 It is calculated from each rule. The production score can also be deduced 296 from a classifier. Each branch (hypothesis) is characterized by a score. 297

The breadth-first exploration using a local context of the IMISketch method is illustrated in Figures 4(a) and 4(b). This exploration generates some combinatorics. Consequently, we describe in section 4.2.2.a a new algorithm for constructing analysis trees for reducing the combinatorics.

³⁰² 4.2.2.a. Our new IMISketch+ method of tree construction

Each analysis tree characterizes the elements to interpret in the defined local 303 context. Each root is the production rule that would consume this primitive. 304 The number of analysis trees corresponds to the number of possible inter-305 pretations for the current primitive. The analysis tree then contains a set of 306 complete or incomplete objects. An object is called complete if and only if 307 this object can be found in the final result of the document interpretation. 308 For example, in the case of the architectural plans, the complete objects can 309 be walls, doors, windows... An incomplete object is an object that is not 310 complete, but rather a part of complete object, i.e., we do not find it in the 311



(a) IMISketch method at step s



(b) IMISketch method at step s+1



(c) IMISketch+ at step s. The direct son of the root are grouped by consumed elements



(d) IMISketch+ at step s+1. Only the roots (n1, n2), belonging to the same group (Group 1) are developed

16

Figure 4: IMISketch Vs IMISketch+: transition between two successive analysis trees. Nodes modeled by circles refer to complete object. Nodes modeled by triangles refer to incomplete objects. The grayed leaves indicate the new applied production rules after movement of the local context. final result of the interpretation. These objects can lead to complete objects(furniture, openings).

Despite the use of a local context, the problem of combinatory and therefore computing time, is an important problem that must be addressed to meet the criteria of acceptability and usability of this type of interactive analysis with a user. The combinatory is mainly due to the number of applied production rules in the local context.

In fact, the development of certain nodes in tree analysis is useless. That is why, we present a new approach to build the tree analysis in order to only develop the useful nodes for making the right decision [16]. This algorithm is described below:

• If the number of analysis tree roots is equal to 1: (Figure 4(c))

- 1. limit the development of direct sons of the root. If the root to de-324 velop is unique, we can say that only one interpretation is possible 325 for the consumed element. The decision process is then sure to 326 validate the right root. In this case, since we know the root to be 327 validated, we consider unnecessary to build all the tree analysis, 328 and we can limit the tree development to the direct sons. After 329 validation of this root, these direct sons will be the new roots to 330 build. 331
- If the number of analysis tree roots is greater than 1: (Figure 4(d))
- 333 334
- 1. regroup the roots by consumed elements. We will not build all roots but only those that share the same elements. We want

335	to find the right interpretation in order to consume the element
336	within the root.
337	2. order these groups by their scores. Each group has a score de-
338	rived from the roots within it. This score is the score of the best
339	hypothesis (branch) located in the group.
340	3. develop only roots belonging to the group having the best score.
341	4. build analysis tree as long as the following conditions are verified:
342	- the newly consumed element is in the local context of research;
343	- the number of consumed elements in each hypothesis (branch)
344	is below a threshold. In the architectural plans, we note that
345	the number of primitive of an opening can not exceed 10 prim-
346	itives. For this we fix this threshold to 10.
347	- the number of complete elements belonging to a branch is less
348	than a second threshold. We can fix a threshold to 3 for the
349	architectural plans.

Figures 4(d) and 4(c) illustrate an example of building the analysis trees with our new method (IMISketch+). The local context is not only limited to the distance between primitives but also to the number of complete elements in hypothesis. This optimization can generate a lack of information on hypotheses and therefore a possible ambiguity. But thanks to the interactivity, this insufficiency does not influence on the final result as far as the user may be solicited to validate the right hypothesis.

357 4.2.2.b. IMISketch+ Vs IMISketch

The aim of this section is to show the improvements deduced from the new 358 algorithm (IMISketch+) for building analysis trees compared to the classic 359 exploration method IMISketch (in which all possible branches are explored). 360 To facilitate this comparison, we present an example of an artificial tree 361 analysis. The goal is to compare, for each step, the number of developed 362 nodes according to the placement of the local context depending on the 363 elements to be analyzed. The transition of an analysis tree to the next 364 analysis tree involves a shift of the local context. This movement allows 365 applying the other productions. In fact, the building of the new analysis 366 trees is not the rebuilding of the whole branches but only the new found 367 productions by shifting the context. 368

Figure 4(a) shows the result of the exploration according to the IMISketch 369 method. The number of interpretation (nodes) is equal to 80 interpretations. 370 The tree construction based on the IMISketch+ algorithm generates only 6 371 interpretations (Figure 4(c)). Figures 4(b) et 4(d) illustrate the new analysis 372 trees by moving the local context. After two successive construction steps 373 of analysis trees, we went from 111 interpretations (IMISketch method) to 374 28 interpretations (IMISketch+), this corresponds to a gain of about 75% of 375 computation time. These optimizations are generic and do not depend on 376 the category of structured document to analyze. 377

378 4.2.3. Making the decision

Once the tree is well constructed, we start the decision phase. The role of the decision process is to validate the right hypothesis among a set of competing hypotheses generated with a descending breadth first analysis. Two cases are presented (section 4.2.2): case of a single root and case of several roots.

• Case of a single root: this root is validated.

- Case of several roots: the hypothesis (branch) having the highest leaf
 score in each tree is considered. These branches are sorted by the score
 of each leaf. Then we compare the obtained branches. Two cases may
 appear:
- Implicit validation: when the analyzer is confident enough to
 choose the correct root without requesting the user. The analyzer implicitly validates the root which has the branch having
 the highest score.
- Explicit validation: when the decision process requires the user to validate the right decision. In practice, if the difference between the branch with the highest score and another branch is below a threshold, called *threshold of confidence* and these two branches are contradictory (at least one joint primitive is not consumed by the same rule production), the user intervention is required.

The decision is not limited only to validate the correct root but can also validate directly a part of the branch (hypothesis), for accelerating the analysis. In general, if the direct son of a node is unique, the validation of this node generates automatically the validation of its direct son. The decision making is summarize in Algorithm 1.



Algorithm 1: Decision algorithm

When the correct root is validated, other roots are put on hold and the new roots are either the sons of this root if exists, or the waiting roots otherwise and the analyzer goes back to the first step (defining the local context step). The analysis is complete when no more production rule is applicable.

409 5. Implementation of IMISketch

In this section, we describe the implementation of our interactive analysis method (IMISketch) and illustrate it on 2D handwritten architectural plans.

412 5.1. Use of CD-CMG

Given that architectural plans are two-dimensional structured documents, 413 for modeling our interactive analyzer, we adopt the two-dimensional gram-414 mars. We use the context-driven constraint multiset grammars (CD-CMG) 415 (detailed in [28]). The analysis process is driven by the context that involves 416 a significant decrease in the complexity of the analysis process. The score 417 calculated by each production (node) is due to preconditions and constraints 418 of the rule production [28]. Equation 1 defines the manner that we calculate 419 the score for each production. The use of the square root is a normalization 420 using a geometric average. 421

$$\rho_P = \sqrt{\mu_{preconditions} \cdot \mu_{constraints}} \tag{1}$$

Deducing $\mu_{preconditions}$: the computing of $\mu_{preconditions}$ is simply the fuzzy application of the precondition block. Each DSC ² is evaluated and the corresponding membership degrees are merged: a fuzzy conjunction (t-norm) is used for an '&', whereas a fuzzy disjunction (t-conorm) is used for an '|'. Once again, the resulting degree is normalized to avoid giving an advantage to productions with less DSC.

Deducing $\mu_{constraints}$: the computing of $\mu_{constraints}$ is based on the same principle than the computing of $\mu_{preconditions}$. It means that each constraint (structural or statistical) must return a membership degree. In practice, defining fuzzy structural constraints is often straightforward and recognition systems we use are based on fuzzy inference systems from which we obtain

 $^{^{2}}$ DSC is a specific constraint modeling both a location in the document and elements that are awaited in it.

⁴³³ such degrees [34].

Equation 2 determines the degree of adequacy (score) of a hypothesis. (PS) is the number of production in the considered branch (referred PS). ρ_{P_i} is calculated by Equation 1.

$$\rho_{PS} = \left(\prod_{P_i \in PS} \rho_{P_i}\right)^{\frac{1}{|PS|}} \tag{2}$$

A production rule can call an external classifier to recognize the symbols. In our application context, the classifier is used for the recognition of the types of opening (eg door, window, etc.). Each recognition is associated to a score. This classification system is based on first-order Takagi-Sugeno (TS) fuzzy inference system [1].

442 5.2. Impact of the primitives

To demonstrate the interest of using the polygons (cf. section 3) to reduce the combinatorics, we compare a first set of production rules for interpreting primitives composed only with segments (Table 1(Ps-P5)) and a second set of rules dealing with two kinds of primitives : the segments and polygons. This second set consists of the same rules enriched by three new rules that allow to analyze polygons (Table 1(P6-P8)).

By applying the production rules described in Table 1 to a set of segments extracted from an architectural plan (Figure 5(a)), we obtain the analysis tree illustrated in Figure 5(c). In the next step, the local context is shifted (Figure 5(b)). We then obtain tree analysis described in Figure 5(d). Despite this example is quite simple, these two analysis trees contain 28 nodes.

⁴⁵⁴ Now, we integrate the new polygon primitive. We add production rules ⁴⁵⁵ illustrated in Table 1(P6-P8). To interpret the segment 1, the analyzer de-

Name	Created elements	Consumed elements
Ps	Starting	longest gogmont in the degument
	wall	longest segment in the document
P1	Wall	segment at the end of an other <i>wall</i>
P2	Sequence	segment at the end of an interpreted wall or sequence
P3	Opening	a sequence and two collinear walls or
	and wall	a <i>sequence</i> and <i>wall</i> and a segment which are collinear
P4	Door	an opening
P5	Window	an opening

Table 1: Example of production rules used for the architectural plans using **segments** (Ps-P5) and using **polygons**(P6-P8)

Name	Created elements	Consumed elements
P6	Wall	polygon at the end of an other <i>wall</i>
Ρ7	Sequence	polygon at the end of an interpreted <i>wall</i> or <i>sequence</i>
P8	Opening	a sequence and wall and a polygon which are collinear
	and wall	



(d) Analysis tree at step n+1. The grayed leaves indicate the new applied production rules after shifting the local context.

Figure 5: Position of the local context (box) during the analysis and the associated analysis trees. The primitives only consist of segments.

termines the local context (Figure 6(a)) and builds the tree analysis shown in Figure 6(c). In the next step, the segment 2 will be interpreted, the local context associated with this segment is illustrated in Figure 6(b) and the new tree analysis is described in Figure 6(d). Thanks to the chaining of segments in polygons, we can reduce the combinatorics to 50%. We went from 28 nodes to 14 nodes.

We have shown in this section, the impact of using the chaining of segments on building of the analysis tree. This improvement induces optimizations in the decision process. Moreover, using polygons increases the chances to validate a set of nodes. (cf. section 4.2.3).

The use of polygons as a primitive not only reduces the combinatorics 466 for structural recognition but also guarantees a better shape recognition for 467 obtained symbols through a more accurate representation of its constituents. 468 Indeed, once the symbol is well recognized structurally, all the segments 469 forming the symbol (the segments that are primitive and chained segments 470 forming the polygon) will be forwarded to the classifier to label the symbol. 471 Indeed, once the symbol is well structurally recognized, all the segments of 472 symbol (the segments that are primitive and chained segments forming the 473 polygon) will be transmitted to the classifier to label this symbol. 474

Once the tree analysis is explored, the analyzer starts the final phase: the decision making. In this phase, illustrated in the next section, the user may be solicited for removing ambiguities.

478 5.3. Ambiguity cases

In this section, we show an example of ambiguity which requires a user solicitation. The aim is to analyze the set of primitives extracted from an



(a) Local context at step n (centered (b) Local context at step n+1 (cenon the segment 1)
tered on the segment 2)



(d) Analysis tree at step n+1. The grayed leaves indicate the new applied production rules after shifting the local context.

Figure 6: Position of the local context (box) during the analysis and the associated analysis trees. The primitives consist of segments and polygons. The dotted line (primitive 3) means the polygon.

architectural plan (Figure 7(a)). At the step illustrated in Figure 7(b), the 481 decision process decides to report the decision on the user because the two 482 branches in competition are contradictory and the difference between the 483 two scores is lower than the threshold of confidence. The decision process 484 launches an interaction with the user (Figure 8), and proposes to validate the 485 first hypothesis. If the user validates this hypothesis, the document will be 486 recognized as two doors (Figure 7(c)). If the user declines the first hypothesis, 487 the system implicitly validates the second hypothesis (Figure 7(d)) and the 488 interpreted document will be made of one window. We note that the interface 489 is basic for the first version; we have not focused on the HCI design but on 490 the fact to interact with user. We focus on "when" the interpretation process 491 solicits the user. "How" to interact will be the subject of our future work. 492

493 6. Experimental results

In this section we report different results obtained with the complete 494 interactive recognition system. The results are focused on the impact of the 495 presented optimizations : the use of the polygon primitive and the building 496 strategy of tree analysis in terms of recognition rate and of computing time. 497 For this reason, we propose three versions of IMISketch (Figure 10). The first 498 one (referred as IMISketch) explores all the hypotheses (branches) of the tree 499 analysis in a local context with a set of primitives that contains only segments 500 (without polygons). The second version (referred as IMISketch+(Seg)) takes 501 into account the new algorithm to build the analysis tree keeping the same 502 set of primitives (only segments) as in the first version. The last version 503 (referred as IMISketch+(Seg&poly)) incorporates the both optimizations: 504



(c) The first hypothesis: transforming a set of primitives into a wall and two doors

(d) The second hypothesis: transforming a set of primitives into a wall and a window

Figure 7: Example of ambiguity between two hypotheses



Figure 8: User interface: the analyzer proposes a user to validate the hypothesis described in Figure 7(c). If the user don't validate this hypothesis, the analyzer will validate the hypothesis implicitly described in Figure 7(d)

the use of the polygon primitive and the new algorithm to build the analysis tree (Figure 10).

The experiments were carried on 69 handwritten architectural plans of 507 varying complexity drawn by around ten different people. Each architec-508 tural plan is composed of a set of walls, doors, windows and sliding win-509 dows. The databases contains 2641 walls, 555 doors, 377 windows and 401 510 sliding windows. Some examples of architectural plans are illustrated in Fig-511 ure 9(a). For each architectural plan, we compare, for the three versions 512 (IMISketch, IMISketch+(Seg), IMISketch+(Seg&poly)), the final interpre-513 tation result and the obtained computing time (Figure 9). 514

⁵¹⁵ Concerning the recognition rates (cf. Figure 9(b)), we note that the ⁵¹⁶ optimizations performed to IMISketch do not affect system performance. In-⁵¹⁷ deed, we observe a very light difference between IMISketch, IMISketch+(Seg) ⁵¹⁸ and IMISketch+(Seg&poly) in terms of recognition. With the user solicita-⁵¹⁹ tion, during the analysis, we obtain 97.96% with IMISketch, 98.13% with ⁵²⁰ IMISketch+(Seg) and 97.93% with IMISketch+(Seg&poly) (Figure 9(b)).

This can be explained as the hypotheses proposed by the three methods 521 are not the same. IMISketch produces more hypotheses than IMISketch+(Seg)522 and IMISketch+(Seg) generates more hypotheses than IMISketch+(Seg&poly). 523 The number of competing hypotheses is more important in IMISketch than 524 IMISketch+(Seg&poly). This might suggest that there are more chance to 525 have the right hypothesis with IMISketch, but the generated confusions are 526 also potentially more numerous. In the end, the results in terms of recogni-527 tion performance are very comparable. 528

Moreover, the user solicitation decreases from 6 interventions in IMISketch to about 4 interventions in IMISketch+(Seg&poly)(Figure 9(b)). This decrease shows a reduction in the competing hypothesis that can lead to ambiguities. This is due to the polygons which has reduced the number of nodes per branch and therefore the conflicts between hypotheses. The obtained errors (about 2%) are due either to poor calibration of the local context, or the badly drawing of certain symbols.

The details of computation time per image between the three versions 536 (IMISketch, IMISketch+(Seg) and IMISketch+(Seg&poly)) are shown in Fig-537 ure 9(c). The images are ordered by the computing time according to 538 IMISketch+(Seg&poly). We improve the computation time from about 11 539 minutes of average computing time per image (with classical IMISketch) to 540 about 4 minutes 43 seconds (with IMISketch+(Seg)), i.e. a gain of 57%. 541 IMISketch+(Seg&poly) further reduces computation time to attain about 2 542 minutes 31 seconds, which presents a total gain of 77%. Depending on the 543 complexity of the plans, the gain of computing time can reach 90% per image 544



(a) Examples of architectural floor plans

	IMISketch	IMISketch+(Seg)	IMISketch+ (Seg+poly)
Types of primitive	Segments		Segments and polygons
Building the analysis trees	All possible hypotheses in the local context (classical approach)	Using the new approach described in Section 4.2.2	
Number of unrecognized symbols	81	74	82
Recognition rate	97.96%	98.13%	97.93%
Average computation time	11m03	4m43	2m31
Average user solicitation	6.07	5.17	3.59

(b) Characteristics of IMISketch versions and recognition rate on archi-



(c) Computing time per image showing the disappearance of combinatorial peak using IMISketch+ 32



(image with index 67). We note, in Figure 9(c), that the peaks in the analysis 545 time have completely disappeared with the optimized method. Thus, there is 546 no risk for a long wait to interpret a more complex plan. These computation 547 times are obtained in real conditions, i.e. in the presence of a user in the 548 loop of analysis. Today, the average computation time is 2 minutes 31 per 549 image. In a future work, we will try to reduce the computation time to attain 550 around 1 min 30 per image: experimentally, it seems the ideal timing that 551 allows the user to follow in real time the recognition process. 552

The experimental results are very encouraging. They suggest that it is 553 possible to introduce a descending breadth first analysis by controlling the 554 generated combinatorics. This supports the idea of conceiving an interac-555 tive systems for the document recognition. User solicitations, driven by the 556 analyzer, guarantee the obtaining of very high rate of reliability even when 557 considering the treatment of complex documents. the use of the polygon 558 primitive does not have a negative impact on the structural recognition rate, 559 in addition it reduces the number of user intervention during the analysis 560 and also accelerates the computation time. 561

⁵⁶² Current state of IMISketch system

Optimizations of our IMISketch method were carried out to the interpretation of more complex architectural plans containing a dozen furniture types (toilet, table, bed...), 3 types for openings (door, window and sliding window) and walls. Indeed, some experiments on 24 architectural handwritten plans (see Figure 10) show that the structural recognition rate increase from 91% without user solicitations to 96% with user solicitations.

⁵⁶⁹ We can notice that the best compromise (recognition rates/user solicita-



Figure 10: Interpretation of a complex architectural floor plans

tions) is obtained with 12 user solicitations per image: it means that 4% of primitive interpretations require the user solicitation. 49% of user solicitations are useful to take the right decision which is not the best hypothesis proposed by the analyser (more details in [16]).

574 7. Conclusion

In this paper, we have presented a complete generic method to interpret 575 sketches such as 2D architectural floor plans. This method consists of a 576 preprocessing phase in which we extract useful primitives which constitute 577 the inputs of an interactive analyzer. This analyzer is based on a competi-578 tive breadth-first exploration of the analysis tree according to a dynamical 579 local context of the document. The decision process is able to solicit the 580 user in the case of strong ambiguity. Generally, the competitive breadth-first 581 exploration generates combinatorics. Our interactive method integrates an 582 optimization strategy for solving combinatorics. This strategy concerns the 583

preprocessing phase and the analysis phase. The preprocessing phase avoids 584 the aggomeration of primitives in reduced zones of document. At the anal-585 ysis phase, the strategy is based on a dynamic construction of analysis tree 586 by controlling the depth of each branch following a set of criteria specified in 587 the current local context. It may be noted that this strategy is generic and 588 therefore it could be easily applied to other types of st uctured documents, 589 and other analyzers characterized by a breadth-first exploration. The first 590 tests of this interactive analyzer have been made on 2D handwritten archi-591 tectural floor plans. Integrating the user in the analysis process is, in our 592 view, a key point to address complex off-line sketch recognition and to avoid 593 an a posteriori verification phase. 594

Future work will focus on extending the experimental results on large image databases containing more complex architectural plans (integration of furniture, quotes, etc.). We will also validate the criteria of acceptability and usability of the system by doing usage tests that will be conducted in collaboration with experts from the Loustic laboratory (http://www.loustic.net/rennes).

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