The Representation and Use of a Visual Lexicon for Automated Graphics Generation

Michelle X. Zhou Steven K. Feiner
Dept. of Computer Science, Columbia University
500 West 120th St., 450 CS Building
New York, NY 10027
{zhou, feiner} @cs.columbia.edu

Abstract*

Most automated graphics generation systems employ either a constructive or a parametric graphics synthesis approach. Constructive graphics synthesis is a deductive approach that builds visual presentations from scratch by gluing together the most basic visual variables. Conversely, parametric graphics synthesis defines a set of parametrized visual models and interprets the information to be presented through instantiation of the selected model. To increase efficiency, we have combined parametric and constructive approaches in a system called IMPROVISE. In this paper, we focus on the parametric aspect of our approach. We present a comprehensive, general, and extensible formalism to represent a visual lexicon for use in automated graphics generation. A visual lexicon is a collection of parametrized primitive visual objects that serve as building blocks for constructing more complex visual presentations. We also illustrate how this representation can be effectively employed to aid the selection and instantiation of a visual lexical item in the graphics generation process. Examples are given from IMPROVISE to demonstrate the representation and use of this visual lexicon.

1 Introduction

Automated graphics generation is a computationally complex task. Most research systems that have been developed to explore it use one of two approaches: parametric graphics synthesis and constructive graphics synthesis.

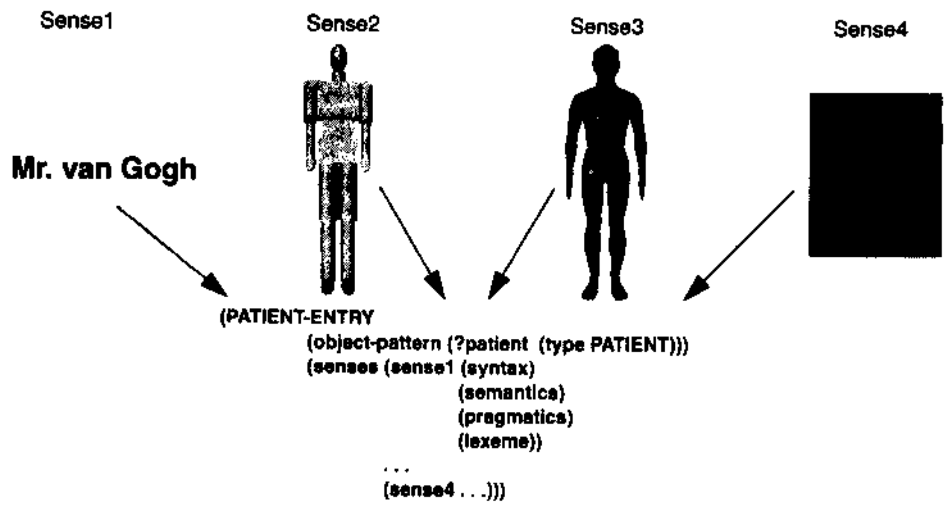
Parametric graphics synthesis defines a set of parametrized visual models. It analyzes the data attributes, maps the data attributes onto the parameters of the models, and instantiates the visual parameters of the selected model to interpret the data (e.g., [Robertson, 1991]). In contrast, *constructive graphics synthesis* does not predefine a set of visual models. Instead, it reasons about visual design principles (e.g., [Mackinlay, 1986]) and attempts to compose visual presentations from scratch by gluing together the most basic visual variables [Bertin, 1983] to form a coherent whole.

We are developing a system called IMPROVISE (Illustrative Metaphor Production in Reactive Object-oriented VISual Environments) that can automatically generate visual presentations using a combination of parametric and constructive approaches. Unlike most of the constructive systems (e.g., [Mackinlay, 1986; Roth and Mattis, 1991]), IMPROVISE constructs a wide variety of visual presentations, including 2D static graphs and 3D animations, to convey heterogeneous information [Zhou and Feiner, 1996]. IMPROVISE begins by sketching abstract plans from scratch to create partially specified visual presentations. These vague descriptions are progressively replaced by more complete descriptions [Zhou and Feiner, 1997]. In a top-down fashion, a complex visual presentation is recursively built by using primitive visual objects that can be handled by a rendering component. We refer to a collection of the primitive visual objects as a *visual lexicon.* Each primitive visual object in the lexicon is called a *visual word* or a *visual lexical item.*

A visual word can be any type of visual form, ranging from a 2D static text string to a video clip. Rather than list all the possible primitives and their combinations or construct them from scratch, we abstract and parametrize the basic visual forms. We have developed a comprehensive, general, and extensible formalism to represent a visual word. Much like a lexical component in natural language' generation systems [Gates et al., 1991; Elhadad et al., 1997], the representation is not only comprehensive enough to capture the syntactic, semantic and pragmatic features of each word, but is also compact enough for the visual words to be easily manipulated by the design process. Furthermore, the formalism is general enough to support a wide range of visual forms, including 3D graphical models, images, and video. The representation can also be easily extended to accommodate new visual forms.

Based on the lexical representation, we have formulated a set of constraints to guide the selection and instantiation of a visual word. Those constraints come from a wide variety of sources including: syntax (e.g., only certain visual words can be used in a composition), semantics (e.g., only certain visual words can be used to fulfill a particular information-

t This research is supported in part by DARPA Contract DAAL01-94-K-0119, the Columbia University Center for Advanced Technology in High Performance Computing and Communications in Healthcare (funded by the New York State Science and Technology Foundation), the Columbia Center for Telecommunications Research under NSF Grant ECD-88-11111,and ONR Contract N00014-94-1 -0564.



```
(PATIENT-ENTRY
        (object-pattern (?patient (type PATIENT)))
        (senses (sense1 (syntax)
                        (semantics)
                        (pragmatics)
                        (lexeme))
                ...
                (sense4 ...)))
```

Figure 1. Multiple senses for patient entry

seeking goal), and pragmatics (e.g., available resources, such as a type of display, require certain types of visual words).

In the rest of the paper, we first briefly discuss several parametric graphics synthesis systems in Section 2. Then we present our representation formalism for the visual lexicon in Section 3. Based on the representation, we postulate a set of constraints that guide the selection and instantiation of a visual word. Finally, in Section 4 we present our conclusions and discuss possible future work.

2 Related Work

The foundation of parametric graphics synthesis is a set of parametrized visual models. Through careful examination, the mapping between the visual attributes of the models and the data properties can be established in a formal and systematic manner. Most systems that employ a parametric-approach are quite similar but have minor differences in the methodologies they use to map data onto their visual models. Nevertheless, systems can be differentiated by the richness and flexibility of their parametrized visual models, as well as their capabilities to compose various models into a coherent presentation.

BHARAT [Gnanamgari, 1981] was developed at University of Pennsylvania to generate 2D displays for quantitative information: pie graphs, bar graphs, and line graphs. It maps numeric data to one of these graphs determined by the data characteristics. Once a graph type is selected, it is instantiated based on the data properties and its syntactic parameters.

Robertson's natural scene paradigm (NSP) [Robertson, 1991] maps data variables to various features of a 3D realistic natural scene, such as surface condition, surface height, and density based on data characteristics. He postulates a set of mapping rules to correlate the properties of the natural scene, the data characteristics, and the user's interpretation goals. NSP focuses on representing scientific or statistical data sets, it does not address how to represent qualitative (i.e., non-quantitative) information.

VISTA (Senay and Ignatius, 1994] is a knowledge-based visualization system that suggests various visual techniques for a given data set and also allows the user to modify the design interactively. Compared to other parametric systems, VISTA offers a much richer set of visual techniques that can effectively express a wide range of scientific data. Furthermore, it employs a sophisticated compositional design approach to design a composite visual display. VISTA partitions a large data set into simple sets of data, selects unused primitive visualization techniques to represent each small data set, and assembles all selected primitive visualization techniques together. Like NSP, VISTA also emphasizes scientific visualization and does not address visual techniques for representing qualitative information.

3 Visual Lexicon Representation and Usage

Since most parametric synthesis systems have focused on presenting quantitative information, their representation formalisms usually are not comprehensive or general enough to handle heterogeneous information. Moreover, most of their parametrized visual models are characterized by their syntactic features: the semantics and pragmatics of the models are cither ignored or simplified. However, to effectively synthesize visual presentations, we need a comprehensive representation of each model that captures its syntactic, semantic, and pragmatic features. The representation formalism should also be general and extensible. In our case, each visual word in the lexicon corresponds to such a model. To meet these criteria of comprehensiveness, generality, and extensibility, we have adopted an object-oriented paradigm to represent a visual word.

3.1 Representation

A visual word is always associated with a single domain

```
#Text                            # Simple body model            # Complex body model            # Image
(sensel                          (sense2                        (sense3                         (sense4
  (syntax                          (syntax                        (syntax                         (syntax
    (category VISUAL-UNITY)          (category VISUAL-UNITY)        (category VISUAL-UNITY)         (category VISUAL-UNITY)
    (subcategory TEXT)               (subcategory 3D-SHAPE)         (subcategory 3D-SHAPE)          (media Image TIFF)))
    (media Graphical-Model)))        (media Graphical-File)))       (media Graphical-File)))
```

Figure 2. Syntax for patient entry

object. However, each domain object can be represented by one or more visual words. In a visual lexicon, we say that a domain object has multiple *senses.* For example, an instance of object PATIENT could be displayed as one of the senses shown in Figure 1. Each sense (i.e., visual word) has its *syntax, semantics* and *pragmatics* in the context of graphic design. In addition, a graphical expression, called a *lexeme,* is also stored.

Object Pattern

To map an object onto a visual word, we first need to know the object pattern including the type, attributes, and other object characteristics [Zhou and Feiner, 1996]. As shown in Figure 1, variable ?patient stands for an instance of PATIENT. More complex patterns can also be represented through pattern description keywords. Currently, we have implemented two keywords: *type* and *attribute. Type* specifies the domain of the object, while *attribute* restricts the objects to a subset that satisfies the specified attributive constraints. For example, to represent any PATIENT whose age must be greater than 20, we use:

```
(object-pattern (?patient (type PATIENT) (attribute (> ;age 20))))
```

Note that the name of each attribute is preceded by a ":".

Syntax

Syntax describes the structure or pattern of a visual word. In particular, *category, subcategory,* and *media* together specify the syntactic features (Figure 2).

Category provides the type of a visual word based on the visual hierarchy described in [Zhou and Feiner, 1997], while *subcategory* further classifies the type information. For example, a visual word is an instance of a VISUAL-STRUCTURE. More specifically, it is in the TABLE-CHART subcategory. Although we could go even deeper to further distinguish the visual word types, we have found a two-level hierarchy to be adequate thus far.

IMPROVISE is designed to deal with a wide variety of visual presentation forms (i.e., visual media formats). We allow four media formats: *graphical model, graphical file, image,* and *movie.*

A graphical model describes a visual object in a particular graphics language. In *graphical model* mode, all graphical expressions are explicitly written out as a lexeme. However, in *graphical file* mode the lexeme refers to the name of a file that contains all the graphical expressions. For example, a human model might be expressed in Inventor [Wernecke, 1994J file format in *graphical model* mode as:

```
Human{
    Head (...)
    Body{
        Chest (...)
        Waist {...}...}
}
```

Alternatively, this description can be kept in a file named "human.iv" in *graphical file* mode. *Image* mode signifies that the current visual object is an image file. Optionally, it can be followed by an argument to indicate the image format as shown in *sense4* of Figure 2. The last media category, *movie,* indicates the current visual object is a video clip. Similar to *image,* it can also be further qualified by appending a type argument (e.g., MPEG1).

The advantage of providing media information is to simplify the process of handling multiple media formats and to make the system more general and extensible.

Usually, we specify a set of abstract visual operations (e.g., *Scale,* and *Move)* at a high level without worrying about the implementation details for different media. At a lower level, based on the provided media information, the abstract operations are realized by media-specific procedures. Hiding the details of abstract visual operations makes it easier to extend the system. Consider the abstract visual operation *Scale. Scale* transforms the modeling matrix for a graphical model, which can change the dimensions of an image. To support a new media format, we only need to write a set of procedures to perform the *Scale* operation for the new medium without affecting the rest of the knowledge base.

Semantics

While syntax focuses on describing the structure or pattern of a visual word, semantics abstracts the meaning of the syntactic features, summarizes the thematic roles of a visual word, and identifies the scope of its role. Figure 3 shows the semantic features for the PATIENT-ENTRY.

Semantic *sense* specifies the abstraction of syntactic features of a visual word, while semantic *role* indicates what a visual word is capable of when it participates in a visual presentation. Moreover, semantic *scope* elaborates how well a

```
# Text                           # Simple body model            # Complex body model            # Image
(sensel                          (sense2                        (sense3                         (sense4
  (semantics                       (semantics                     (semantics                      (semantics
    (role Identify)                  (role Locate)                  (role Locate)                   (role Identify)
    (scope Naming)                   (scope SpatialPosition)   (Scope   SpatialPosition)            (scope Appearance)
    (sense LABEL)))                  (sense SYMBOL)))               (sense SYMBOL)))                (sense PORTRAIT)))
```

Figure 3. Semantics for patient entry

```
# Text
(sense1
  (pragmatics
    (domalnlnfo ...)
    (hardware
       (platform SGI I PC)
       (cpu R4400 I Pentium...)
       (graphics Extreme I...)
       (display Color I Grayscale))
    (performance  FAST))...)
```

```
# Simple body model
(sense2
  (pragmatics
    (domaininfo...)
    (hardware
       (platform SGI I PC)
       (display Color I Grayscale))
    (performance FAST))...)
```

```
# Complex body model
(sense3
  (pragmatics
(domaininfo             ...)
    (hardware
       (platform SGI)
       (display Color I Grayscale))
    (performance MEDIUM))...)
```

```
# Image
(aense4
  (pragmatics
    (domaininfo...)
    (hardware
       (platform SGI I PC)
       (display Color I Grayscale))
    (performance  MEDIUM))...)
```

Figure 4. Pragmatics for patient entry

visual word can perform in the specified role. In Figure 3, *sensel* is a *label* that can only *identify* the patient by *naming,* but *sense4* is a *portrait* that can *identify* the physical *appearance* of the patient.

Pragmatics

In language or image understanding, pragmatic features usually refer to the roles or influences that the language or image possesses to help the user understand distinguishing features and appropriate contexts (e.g., [Goldsmith, 1984]). To facilitate visual design, we have extended the connotation of the pragmatic features of a visual word. Not only do we include word features that are directly related to the characteristics of the user and context, such as the user's identity, expectations, and application type, but we also cover factors that describe the relationships between the word and the design or execution environment, such as the hardware requirements (e.g., display type) of the word. Since modeling users and system environments is a complex task itself, we concentrate on a few types of information that are particularly useful. They are *domain information, hardware requirements,* and *performance estimation.* Figure 4 lists the pragmatics of each word in PATIENT-ENTRY, with *domain information* omitted since they all share the same one:

```
(domalnlnfo
        (appType  MEDICAL)
        (audienceType NURSE I DOCTOR))
```

The pragmatics features of *sensel* in Figure 4 can be read as: the text representation of a patient's name is suitable for both nurses or doctors in a medical application. Moreover, this text can be rendered on a SGI or PC with color or grayscale display. The rendering speed of such a text string is fast (e.g., under a millisecond).

Although the pragmatic features involve domain-specific information, we can consolidate all domain-specific information by expressing it in a disjunctive form. For example, if we want to use the PATIENT-ENTRY for both medical and military logistics applications, we can write:

```
(domalnlnfo
        (infol  (appType MEDICAL)
                (audienceType  NURSE  I  DOCTOR))
        (Info2  (appType  LOGISTICS)
                (audienceType  MEDIC  I  DOCTOR)))
```

Lexeme

In a visual lexicon, each lexeme is a graphical representation of a visual word. Each type of representation *{graphical model, graphical file, image,* and *movie)* is actually a parametrized template. For example, the lexeme of *sense1* in PATIENT-ENTRY is:

```
(lexeme  (VISUAL-UNITY  (geometry  (Text2
                                    (string (get-name ?patient))))))
```

This says that the visual word is an instance of VISUAL-UNITY with a geometry of 2D text, which in turn requires that the string be the name of a patient. Embedding procedures in the representation can be very useful. It eases information encoding by abstracting the common features of a visual word and expressing them in a procedural format. Suppose that every patient's picture is stored as an image in the knowledge base, and is named by the patient's ID number with a suffix "tif'. Instead of explicitly listing all the patients and the file names of their pictures in the visual lexicon, we can use a single expression:

```
(lexeme (IMAGE (fileName (string-cat (get-id ?patient) ".tif"))))
```

This states that the file name can be constructed by concatenating the patient's ID and the suffix ".tif"

To be consistent with the constructive theme of IMPROVISE, the visual lexicon stores only visual words that represent atomic objects. Visual representations for composite objects are built from scratch by piecing together their component representations.

3.2 Usage

IMPROVISE'S visual lexicon is first searched to select a visual word for a given object and then the parameters of the selected visual word are instantiated.

Visual Lexicon Selection

As described above, only atomic objects can be directly matched against the *object-pattern* specified in a visual lexicon entry. For example, suppose the task is to display patient Jones's demographics information to a nurse in a medical application. In our case, "demographics" is a composite object whose component objects include name, age, and gender. The system iteratively refines the abstract plan for accomplishing the task and reaches the point where it must decide how to represent the "age" object. At this point, the system needs to match the object description (i.e., Jones-Age) against each *object-pattern* in the lexicon. The pattern matching is conducted by unifying the object description with an *object-pattern.* If the unification is successful, then all the visual words for the specific *object-pattern* become candidates, of which one will be selected to represent the object. In this example, the two unified entries are:

```
Object:     (Jones-Age (type PHYSICAL-ATTRIBUTE))
```

Pattern-1: (?attribute (type ATTRIBUTE))

Note that the two matched types are not necessarily the same. But PHYSICAL-ATTRIBUTE is indeed a subclass of ATTRIBUTE. For generality, matching by inheritance is allowed. If an object description matches more than one *object-pattern,* the system always chooses the most specific one. The lower the object type is positioned in the domain ontology hierarchy, the more specific the object type is. Moreover, the more attributive constraints an object pattern has, the more specific the pattern is. Naturally, the more specific object pattern inherits the lexemes that are defined for more general object patterns. Suppose there is another *object-pattern:*

Pattern-2: (?attribute (type PHYSICAL-ATTRIBUTE))

In this case, *pattern-2* is used instead *of pattern-1* and *patter-2* also inherits all lexeme expressions defined in *pattern-1*. Moreover, through unification, the variable specified in the *object-pattern* is bound to the matched object. Thus, variable ?attribute is bound to object Jones-Age.

The task of the selection stage is to single out the most appropriate candidate. Many factors can affect the decision. We have established a set of *syntactic, semantic,* and *pragmatic* constraints to aid the selection.

Syntactic constraints. Syntax governs the pattern formation in a synthesis process [Marks, 1991]. Syntactic features specified at different visual presentation levels are used to filter out undesired patterns to keep them from participating in higher-level patterns. Figure 5 was automatically generated by IMPROVISE to present a patient's information summary to a nurse. The top-level visual description for the entire display is a STRUCTURE-DIAGRAM [Lohse et al., 1994], which is a subtype of VISUAL-STRUCTURE. The syntax of a STRUCTURE-DIAGRAM requires that its *core* component be a VISUAL-UNITY that is *not* in the TEXT subcategory. In this case, the *core* component could be one of the patient's visual representations shown in Figure 1.

Based on the descriptions in Figure 2, all visual words satisfy the syntactic constraint except *sense]* which *is* a TEXT. (The patient name in Figure 5 is not generated through defining the *core* component, but as part of the demographics information.) However, there are still multiple candidates

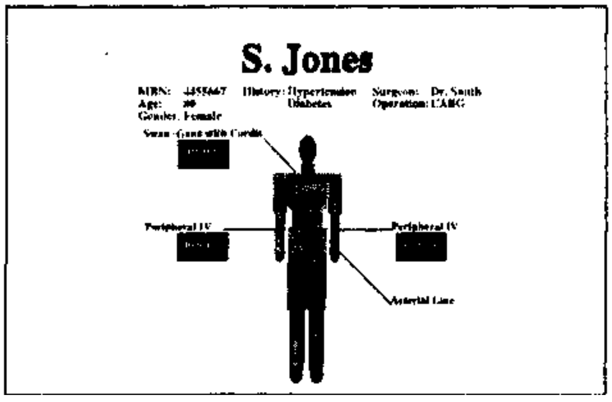


Figure 5. IMPROVISE'S presentation for a nurse

that need to be ruled out.

Semantic Constraints. While syntactic constraints guarantee that visual representations have valid structures, semantic constraints can be used to ensure that the syntactically valid structures are also perceptually correct. Semantic constraints can come from different aspects. One type arises from information-seeking goals [Casner, 1991]. These constraints are evaluated by matching the semantic requirements to the *role* or *scope* description of the visual word. Some visual words can play some *roles* to a certain extent to meet the goals while others cannot. In the example of Figure 5, the task actually requires locating positions on the patient's body so that other information can be arranged around the body. By this criterion, only two choices remain: *sense2* and *sense},* since only these two can locate body positions.

Another type of constraint is specified by high-level visual preferences. These constraints usually require that the selected visual word have a certain visual sense (e.g., be a *symbol* instead of a *portrait),* which in turn defines the potential usability of the visual word [Arens et al., 1993]. For example, a *symbolic* 3D graphical model can be viewed from different angles, while a *portrait* image cannot. In our case, the *core* component of the STRUCTURE-DIAGRAM is required to be a *symbol.* Fortunately, since both *sense2* and *sense3* satisfy the requirements, there is no conflict.

In case all semantic constraints can be satisfied but result in a conflict set, we order the constraints by decreasing importance: first information-seeking goals, then visual preferences. This order in turn determines the importance ranking of the semantic features of a visual word: *role, scope,* and *sense.* In other words, when semantic constraints help determine visual word selection, a visual word should be weighted first to see whether it can fill the *role* to achieve the information-seeking goals, as measured by the *scope* of its role. While not posing any threat for achieving any information-seeking goals, *sense* can be considered last to see whether the user's or application's visual preference can be met.

Pragmatic Constraints. Pragmatic constraints can be used to further filter out some of the undesired candidates. Like other constraints, pragmatic constraints also arise from a wide variety of sources. We have formulated several types of constraints that are most important to our applications. They come from the following aspects: *application situation* (including *presentation type,* targeted *audience,* and *location),* available *hardware, criticality,* and *timing.* Continuing with our patient example, a set of system pragmatic constraints is specified as:

```
(appSituation
        (type SUMMARY)
        (audience NURSE)
        (location CARDIAC-ICU))
(hardware (platform PC)
              (display MEDIUM.RESOLUTION & COLOR))
(timing SHORTJTIME)
(criticality CRITICAL)
```

More specifically, the system needs to present the patient's summary to a cardiac ICU nurse in a short time.

The presentation will run on a PC with a medium-resolution color display. Recall that the candidates left, after screening by both syntactic and semantic constraints, are *sense!* and *sense3.* Based on Figure 4, *sense!* is eventually chosen over *sense3* since *sense!* meets both timing and hardware constraints. This filtering process is done by checking the list of pragmatic features (e.g., performance) of each candidate against the pragmatic constraints (e.g., timing). Based on the success of the match, the candidate set is further pruned.

As in ordering different semantic constraints, we also rank those pragmatic constraints based on their importance: *application situation, criticality, timing,* and *hardware.*

Unlike other graphics generation systems, IMPROVISE does not require a specific order of constraint satisfaction (e.g., satisfying pragmatic constraints before syntactic constraints) in its design process. Instead, IMPROVISE allows the constraints to float with the design process [Elhadad et al., 1997] and solves them when the time is right. If there is enough evidence to indicate that a constraint is satisfiable, then IMPROVISE asserts this fact by taking appropriate actions (e.g., eliminating unwanted candidates). Otherwise, IMPROVISE defers its decision until more information is available. Thus, IMPROVISE operates in a least-commitment manner [Cohen and Feigenbaum, 1989J and avoids unnecessary backtracking over arbitrary decisions.

Visual Lexicon Instantiation

After a visual word is selected, it will be instantiated once there is enough information to supply parameter values. Like selection process, the instantiation also adopts a least-commitment algorithm: no instantiation is made unless there is enough information.

Instantiation involves two steps. The first step is to replace the object variable specified in the *object-pattern.* This is straightforward: the object variable is replaced, wherever it appears, with the name of the object with which the variable is unified. For example, if variable ?patient is bound to patient Jones, an expression such as (get-name ?patient) will be replaced by (get-name Jones). The second step deals with the parameters in a lexeme. For conciseness and efficiency, only key information is explicitly expressed in a lexeme. For example, if text is chosen to represent patient Jones's name, the lexeme in the visual word specifies:

```
(Text2 (string (get-name Jones)))
```

However, this is inadequate for realizing the name: additional information, such as text font or text color, needs to be supplied. Certainly, default values can be provided as part of the knowledge base (as they are in IMPROVISE), but default values are not always the desired ones. Moreover, different visual words have a different set of parameters and the number of parameters can be large (e.g., 10 to 15). Instead of explicitly expressing all parameters for each word in the visual lexicon, a set of variables is automatically generated to represent those parameters once a word is selected. In our example, the expression is actually expanded as:

```
(Text2
    (string (get-name Jones))
    (font?font-1)
    (color ?color-1)
    (justification ?justify-1))
```

Instantiating each variable in this extended format is a challenging task because effectiveness and consistency must be taken into account. Effectiveness constraints from various sources restrict the variable to certain values. For example, if the system runs on a very slow platform, it would probably be better to display a 3D graphical model as a wireframe instead of with shaded polygons. Thus, the parameter *draw-mode* for this particular graphical model might take value *wireframe* instead of *solid.*

Unlike most text generation systems that try to use paraphrasing to avoid repetitions, most graphics generation systems try to maintain design consistency by reusing the same visual cues whenever possible [Zhou and Feiner, 1997]. Unless there is a good reason to do otherwise, similar objects should appear alike in the course of visual presentation. Using variables in visual word descriptions eases this task. For example, if a consistency rule asserts that all text objects that represent a patient's name should appear in the same font and color, then all the variables representing font and color in the text can be co-referred (e.g., ?font-1 is ?font-2, and vice versa). In other words, if any one of them is instantiated, all co-referred variables are instantiated at the same time to the same value.

There are other factors that also need to be taken into account in instantiation; for example, instantiating one variable could affect the instantiation of another. Assuming that the current background color is instantiated to blue, then perceptual rules demand that the text color should not be instantiated to red.

4 Conclusions and Future Work

We have implemented a system that combines a constructive graphics synthesis approach with a parametric graphics synthesis approach. In this paper, we have presented the core component of our parametric approach. A *visual lexicon* is used to generate visual descriptions for atomic domain objects during synthesis. We have described a comprehensive, general, and extensible formalism to represent the lexical entries. Figure 6 summarizes all possible values that are used to describe the syntax and semantics of a visual word. However it is worth noting that some of the values are correlated; for example, if syntactic *category* takes VISUAL-UNITY as its value, then its *subcategory* can never be a TABLE-CHART. Theoretically, we could encode complex visual representations such as TIME-CHART in a visual word. To retain the flexibility and extensibility of constructive graphics synthesis, we usually construct complex visual presentations from scratch while keeping already-made simple visual representations in the visual lexicon. Furthermore, we have formulated a set of constraints to guide the selection and instantiation of the visual lexical items in parametric synthesis.

We believe that this hybrid takes advantage of both parametric and constructive approaches to make graphics generation more powerful, efficient, and flexible. As in a constructive approach, the system is inherently flexible and extensible. As in a parametric approach, we gain efficiency

Syntactic Descriptions

Slots	Value Description	
Category	VISUAL-STRUCTURE	VISUAL-UNITY
	TABLE-CHART	IMAGE
	TIME-CHART	VIDEO
Subcate-	BAR-GRAPH	TEXT
gory	LINE-GRAPH	2D-SHAPE
	PIE-GRAPH	3D-SHAPE
	GRAPHICS-MODEL	GRAPHICS-MODEL
	GRAPHICS-FILE	GRAPHICS-FILE
Medium		IMAGE
		[TIFF I RGB I GIF I PICT]
		VIDEO
		[MPEG1 I SGI I QTIME]

Semantic Descriptions

Slots	Value Description
	LABEL
	LIST
Sense	PLOT
	SYMBOL
	PORTRAIT
	CLUSTER
	IDENTIFY
Role	LOCATE
	DISTINGUISH
	PROPOSITION
Scope	SPATIAL-RELATION
	CONCEPT-FUNCTION

Figure 6. Syntactic and semantic descriptions of a visual word

by providing a reduced search space and facilitate knowledge encoding through reuse of visual words.

Several areas could be further improved to make the approach more systematic and comprehensive. One is to use a good computational model (e.g., a probability model) to describe various constraints more accurately. As most of the constraint values are expressed qualitatively in discrete values (e.g., *short* and *long* describe the timing information), a better model can be used to provide more precise quantitative measurements. Probabilistic reasoning [Russell and Norvig, 1995] could also be incorporated to aid the generation process and model constraints. Last, but not least, the capabilities of each visual word need to be further studied to allow a more comprehensive understanding of various visual forms.

References

Arens, Y., Hovy, E., and Vossers, M. (1993). The knowledge underlying multimedia presentations. In Maybury, M., editor, *Intelligent Multimedia Interfaces,* chapter 12, pages 280-306. AAAI Press/The MIT Press, Menlo Park, CA.

Bertin, J. (1983). *Semiology of Graphics.* Univ. of Wisconsin Press, Madison, WI. (trans, by W.J. Berg).

Casner, S. (1991). A task-analytic approach to the automated design of graphic presentations. *ACM Trans, on Graphics, \0(2):* 11-151.

Cohen, P. and Feigenbaum, E., editors (1989). *The Handbook of Artificial Intelligence,* volume 3, chapter XV. Planning and Problem Solving. Addison-Wesley, Reading, MA.

Elhadad, M., McKeown, K., and Robin, J. (1997). Floating constraints in lexical choice. *J. of Computational Linguistics.* To appear.

Gates, D., McCardell, R., Takeda, K., and Nirenburg, S. (1991). Generation lexicons. In Goodman, K. and Nirenburg, S., editors, *A Case Study in Knowledge-Based Machine Translation,* chapter 6, pages 145-163. Morgan Kaufmann Publishers, Inc.

Gnanamgari, S. (1981). *Information Presentation Through Automatic Graphic Displays.* PhD thesis, Univ. of Penn.

Goldsmith, E. (1984). *Research Into Illustration: An Approach and A Review.* Cambridge University Press, Cambridge.

Lohse, G., Biolsi, K., and Rueter, H. (1994). A classification of visual representations. *Communications of the ACM,* 37(12):36-49.

Mackinlay, J. (1986). Automating the design of graphical presentations of relational information. *ACM Trans, on Graphics,* 5(2): 110-141.

Marks, J. (1991). A formal specification scheme for network diagrams that facilitates automated design. *J. of Visual Languages and Computing,* 2(4):395-414.

Robertson, P. (1991). A methodology for choosing data representations. *IEEE Computer Graphics and Applications,* II(3):56-67.

Roth, S. F. and Mattis, J. (1991). Automating the presentation of information. In *Proc. IEEE Conf on AI Applications,* pages 90-97,

Russell, S. and Norvig, P. (1995). *Artificial Intelligence: A Modern Approach.* Prentice Hall, Englewood Cliffs, NJ 07632.

Senay, H. and Ignatius, E. (1994). A knowledge-based system for visualization design. *IEEE Computer Graphics and Applications,* 14(6):36-47.

Wernecke, J. (1994). *The Inventor Mentor: Programming Object-Oriented 3D graphics with Open Inventor.* Addison Wesley, Reading, MA.

Zhou, M. and Feiner, S. (1996). Data characterization for automatically visualizing heterogeneous information. In *Proc. IEEEInfoVis'96,* pages 13-20, San Francisco, CA.

Zhou, M. and Feiner, S. (1997). Top-down hierarchical planning of coherent visual discourse. In *Proc. IUI'97,* pages 129-136, Orlando, FL.

NEURAL NETWORKS

NEURAL NETWORKS

Neural Nets 1: Rule Extraction