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# **Execution Patterns in Object-Oriented Visualization**

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## Execution Patterns in O b ject-O riented V isualization

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## A bstract.

Execution patterns are a new m etaphor for visualizing execution traces of object-oriented programs. We present an execution pattern view that lets a program mer visualize and explore a program's execution at varied levels of abstraction. The view employs visual, navigational, and analytical techniques that accommodate lengthy, real-world traces. By classifying repetitive behavior automatically into high-order execution patterns, we drastically reduce the information a programmerm ust assimilate, with little loss of insight.

## 1 Introduction

The gap betw een a program 's static specification and its dynam ic behavior is particularly large in object-oriented program s. Experienced developersm igrating from procedural to object technology often complain that they can 't discern flow of control in object-oriented program s. Their complaints grow louder when they deal with one or more fram eworks. The inversion of control from which a fram ework derives its leverage [22] nonetheless gives developers feelings of insecurity, because their understanding of a program is bound tightly to know ledge of control flow. When they can't trace the program 's logic, they can 't predict its behavior, and so they can't change what it does rendering the fram ework useless.

Object-oriented program visualization systems such as Ovation [9,10] and Program Explorer [18] have sought to bridge this gap. A llw ork in much the sameway. They instrument the code to be visualized so that it produces a trace as a side-effect of its execution. The trace contains far toom uch detail (e.g., every method invocation) to understand as is. Instead, the visualization system interprets the trace and generates one ormore animated renditions of the program 's execution. The renditions

use a variety of notations and cognitive devices to convey inform ation succinctly and comprehensibly. The visualization can help hum ansmake sense of millions of method invocations on hundreds of thousands of objects. But to be effective, the visualization system must consolidate and present just the right information on the screen, in just the right way. There are many ways to consolidate information, and their effectiveness varies tremendously.

To understand how to present inform ation effectively, consider how traces become large in the first place. One reason is sheer complexity. A nontrivial object-oriented program may enlist hundreds or even thousands of classes to do its job. One would expect large traces from such programs.

Yetaprogram doesn'thave to be statically complex to yield large traces; even a small program may produce them if it executes the same code many times. Program ming languages let us specify repeated executions of (almost) the same scenario through recursion, functional decomposition, template functions, and a variety of control structures. These remarkably compact specifications produce patterns of similar if not identical instructions as the program executes, greatly lengthening the trace but adding little to its information content. Hence even a simple program can yield huge, incomprehensible traces.

N evertheless, trace data can be comprehended (1) if it can be summarized into succinct, abstract nuggets that lack extraneous detail (details are provided on demand, not unsolicited), and (2) if similar patterns in the trace can be condensed into a smallernum berofmore general patterns that recurat specified frequencies.

To this end we introduce a variation of Jacobson's interaction diagram s [16] for presenting patterns of execution— an execution pattern view—that accomplishes two aims:

 It lets a program m er observe any part of the program 'sexecution atvarious levels of detail. The program m er avoids being overwhelm ed by execution inform ation through careful, se-

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lective control of w hat is divulged, w ith detail presented on dem and.

 It detects and presents generalized patterns of execution in which one pattern subsumes many parts of the trace.

In the next section we describe the elements of the pattern view. We use a scenario to motivate the mechanisms for navigation and detail elaboration. Then we discuss generalizations of execution patterns. We give criteria by which patterns are automatically considered similar and explain our choice of color and shape to best convey the information. Finally, we compare our execution pattern view to related work and offer concluding remarks.

#### 2 Execution Pattern Notation

Jacobson's interaction diagrams are popular because they depict dynamic behavior clearly and compactly. Figure 1(a) shows a simple interaction diagram. Vertical lines, or rails, represent objects. A now headed lines represent message sends: there is an arrow headed line between object A and object B for each message A sends to B, with lines for later messages appearing below those for earlier ones. Thus time progresses downward. All method invocations on an object appear as rectangles superimposed on the corresponding rail.

An interaction diagram 's layoutem phasizes the program 's thread of execution. Before interaction diagram sitw as com m on to depict this inform ation as directed graphs, with nodes representing objects and arcs representing message sends. The problem was scalability. It was difficult to depict more than a few message sends between objects without clutter. By representing time explicitly—that is, by mapping it to the vertical axis—interaction diagram s can depict many more message sends unam biguously.

While interaction diagrams are better than directed graphs at depicting nontrivial interactions, even these diagrams do not scale to complete program executions. Space is consumed quickly as execution time and the number of objects increase. There are ambiguities as well. The order of rails along the horizontal axis is undefined; often it is chosen to minimize line length, crossovers, or both. Nor is it easy to discern the lifetimes of recursive calls, since they are subsumed by a single rectangle. For example, it's unclear in Figure 1 (a) whether the

m essage from A to D is sent within the recursive call to A. One can make recursion more explicit by superimposing rectangles, but that approach quickly gets unwieldy as the recursion deepens.

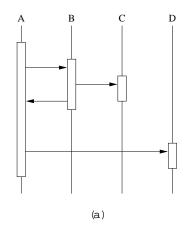
The execution pattern view addresses these shortcom ings by unfolding the graph into tree structures like the one shown in Figure 1(b). This layoutem phasizes the progression of time, not the thread of control. You read the graph left to right and top to bottom, just as you would printed English. Later messages appear further to the right on the same line or further down the screen than earlier ones.

In addition to these topological in provem ents, weuse color to indicate the class of object. Figure 2 shows the execution pattern view at an early point in a program 's execution. The legend in the low er left indicates that objects of class AA are orange, objects of class MM are yellow, and objects of class QQ are blue. To the left of an object is the sender of the m essage it received; to its right, the m essages itsends. Each object in the view is identified with a num erical identifier ("ID") in the upper left of its colored box. M essages are shown as labeled arrow headed lines, the label identifying the nam e of the m essage. In the figure, the m essage fA is sent to an object of class AA (ID = 2). The object responds to this message by sending a message f M to the MM object ( $\mathbb{ID} = 3$ ) and then sending a m essage fQ to the QQ object  $(\mathbb{D} = 4)$ .

A tree representation simplifies things considerably. Its unidirectionality in both axes makes it easier to read than an interaction diagram (where messages may bounce across many rails in either direction). It scales better, too. Horizontal space is mapped not to objects but to the call sequence, which is uniquely defined and more tightly bound—several objects may occupy the same column. Vertical space is used more efficiently as well (compare Figures 1(a) and 1(b)).

## 3 Manipulation and Navigation

The execution pattern view has interactive capabilities that capitalize on the tree structure's properties. Suppose we want to understand how a typical Bus-Observer object in a program handles the update message. The user can initiate a search for execution patterns on various criteria, such as the involvement of a particular class, object, or message name. In this case, the programmer would specify the class Bus-Observer and possibly the



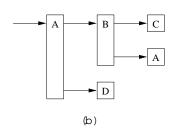


Figure 1: Sim ple interaction diagram (a) and its corresponding execution pattern (b)

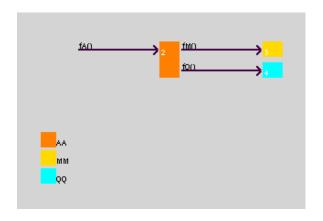


Figure 2: Sim ple execution pattern

update m essage. Then he can brow se through view s, like Figure 3, depicting the w ays different Bus-Observer objects handled the update m essage.

In this example, an initial message update is sent to the black Bus-Observer object ( $\mathbb{D}$ =762). It responds by sending a message notify pending to the purple EClassModel object ( $\mathbb{D}$ =761). Next, the Bus-Observer objects sends a Phrase message to the orange Annobus object ( $\mathbb{D}$ =758). Finally, it sends anotherm essage to the purple EClassModel object ( $\mathbb{D}$ =761).

## 3.1 Collapsing and Expanding Subtrees

Now suppose the program mer wants to explore the response of this update message to the Bus-Observer object (ID = 762) in more detail than Fig-

ure 3 provides. You can see that the depiction of the Annobus object (ID = 758) and the low er (i.e., later) depiction of the EClassModel object (ID = 761) have a beveled border, making them look raised as opposed to flat. A raised rectangle indicates that the object reacted to the stimulus by sending one ormore messages.

Clicking on a raised object reveals them essage(s) that this object sent along with the object(s) that received the message(s). After clicking on subsequent raised objects, we get a view like the one in Figure 4. (Note the self-invocation of the EClass-Model object.) All objects appear flat now, meaning that no hidden messages remain. If we don't want to see a part of the execution, we can collapse part of the view by clicking on a flat object, thereby hiding its responses. The object will now appear beveled as before.

This sim ple technique of expanding and collapsing is a helpful navigation tool. The program mer can selectively drill down to any level of detail without being flooded with information. Moreover, them etaphor is reminiscent of how encapsulation works in object-oriented programs: details of how an object performs a given task are hidden unless sought explicitly.

## 3.2 Changing Context

A program meris likely to ask two questions at this point: "Who sent the initial message update to Bus-Observer 762?" and "What was the context of that message?" The system can take us up a level to view the sender of this update message.

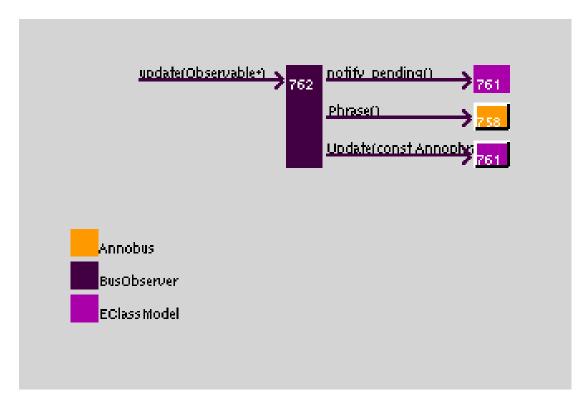


Figure 3: M essage update sent to Bus-Observer object, and its response

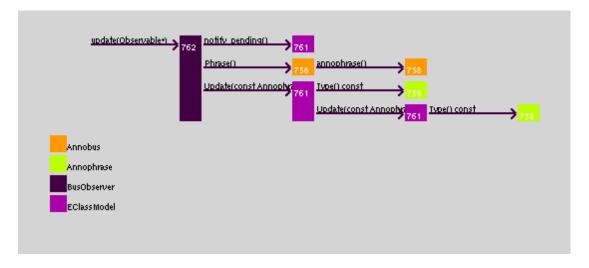


Figure 4: M essage update sentto Bus-Observer object and its fully expanded response

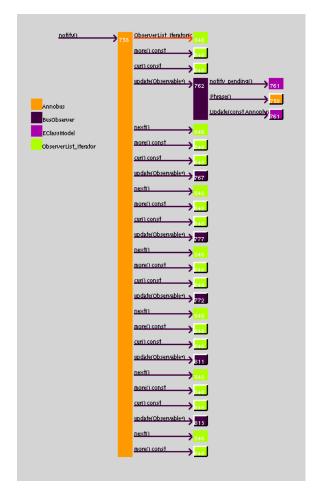


Figure 5: Revealing the sender of the update message: an Annobus object

The result is the view in Figure 5, which reveals that Annobus object 758 sent the update message to Bus-Observer 762. Annobus 758 also sent messages to several other objects in response to a notify message it received.

## 3.3 Filtered Expansion

W hatifthe userw ants to see every message a Bus-Observer receives, filtering out as many other interactions as possible? That's done by expanding only those nodes in the tree that lead to a Bus-Observer object, collapsing all subtrees not containing Bus-Observer. We call this filtered expansion.

Filtered expansion can be useful for identifying patterns that entail object instantiation. Object-oriented programs do nothing if they don't create

objects, and understanding how they create them is often more in portant than knowing when they do. To track down such patterns, we can filter out all messages that do not ultimately lead to a creating (and destroying) message.

## 3.4 Repetition

Figure 5 shows considerable repetition in the execution pattern: a series of four patterns (initiated by more, cur, update, and next) is repeated six times. Indeed, if we were to expand the black and green colored objects, we would find that the collapsed portions exhibit the same pattern, too. While this example has a repetition factor of only six, often repetition factors are much higher. Visualizations of such repetitive sequences rarely pull their weight: they take a bt of screen space without adding much information.

The system can autom atically detect repetitions in patterns; the result is shown in Figure 6. The sequence that was repeated six times now appears raised, indicating the repetition factor in its lower left corner. This compact representation corresponds to a loop in the source code.

Further, just as loops can be nested in source code, the system detects and represents loops at any level of nesting. If we navigate up a few times, we get a view that looks like Figure 7. The innermost frame indicates six repetitions, the middle frame two, and the outermost frame 31. Obviously, this pattern would take a lotmore space if the repetitions were expanded. Objects within the raised frame may be expanded and collapsed as before: note how some objects in Figure 7 are flat (expanded) and some are raised (collapsed).

This exam ple show sa repetition of an iteration sequence. Another important kind of repetition is a recursion sequence. While iterative patterns repeat in the vertical direction, recursion shows up as repetition in the horizontal direction. We can apply the same techniques to make recursion more compact.

## 3.5 Zoom ing and Panning

The system supports zoom ing and panning, two traditional perspective-changing mechanisms. Figure 8 shows a zoom ed-outview of the highest level of the program. A lthough we are using the same pattern view, zoom ing serves other purposes than those we'ved iscussed. At this mag-

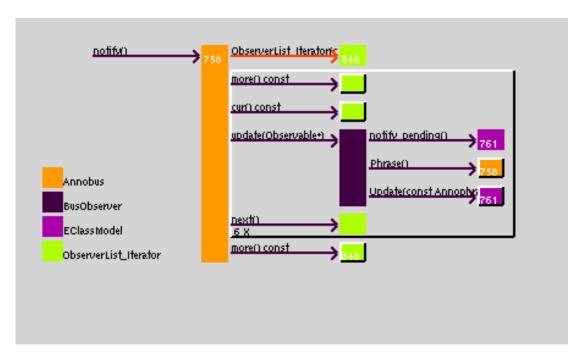


Figure 6: Raised fram e indicates repetition (6X) of the sequence inside

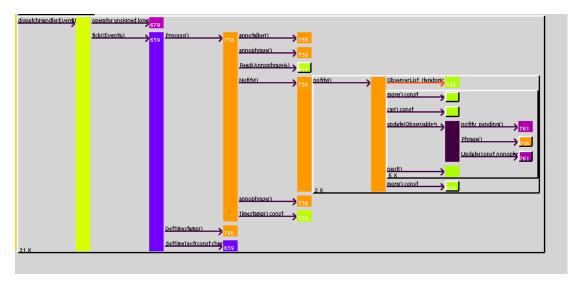


Figure 7: Three nested repetitions

nification level it's im possible to discern individual messages or objects. (In fact, the view om its text and other minutiae when they are too small to read.) What we get here is a general idea of the different phases in the program. The dominant colors indicate the classes that are prevalent in each phase. Note also that the width of the pattern reflects the stack depth at a particular phase of the program.

## 3.6 Flattening and Underlaying

Som etim es even the stack depth gets overwhelm ing. In that case you might have to ignore certain objects, classes, messages, or combinations thereof. You might decide to ignore library classes, for example, or private methods. Filtered expansion (Section 3.3) doesn't help you here. A collapsed node encapsulates the entire response to a message send. Once collapsed, none of the details are visible.

The execution pattern view lets a node encapsulate a specific outgoing message by collapsing only the receiver of the message. We call this flattening the receiver. Both the message and the receiver's identity are hidden; how ever, the fact that a receiver exists and its type are still discernible.

Figure 9 illustrates different flattening com binations. In each case only an indication of the receiver rem ains after flattening—just enough to reveal its existence and color (class) but not its label (identity) or the arrow headed line (m essage).

Another elision mechanism removes a node from the graph even as it highlights the node's encapsulation of its children. Figure 10 shows a node underlaying the subtree itencapsulates. The messages sent by the underlaying node are hidden, thereby reducing visual clutter and saving horizontal space.

Collapsing, flattening, and underlaying differ in their elision properties. Collapsing achieves the greatest elision and hence the greatest space savings, but it leaves no clues about the patterns it elides. Flattening retains type information in exchange for only slightly more real estate, but it hides object identity. The space savings are least with underlaying, but it simplifies the view with minimal information loss: only a few messages are hidden.

The execution pattern view lets you m ix and m atch these techniques to exploit their strengths. For example, a node underlaying a flattened sub-

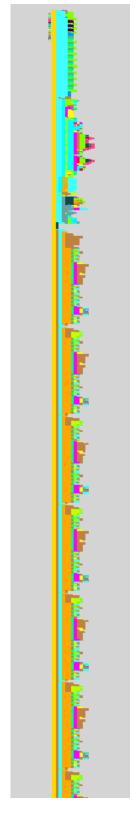


Figure 8: Zoom ed-outview of the entire execution

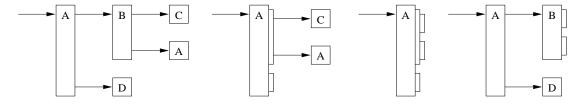


Figure 9: Schematic view offlattening

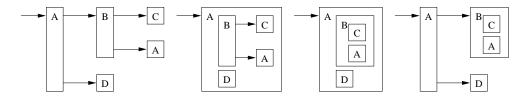


Figure 10: Schematic view of underlaying

tree can be collapsed. In addition, a user can specify elisions for the system to use on a per-class basis. (By default, the system automatically flattens associative classes (Section 43.7), and it underlays metaclasses in Smalltalk and standard library classes in C++.) Users can distinguish these classes at a glance, making the diagrams easier to assimilate and interpret.

### 3.7 Charts as Subtrees

Collapsing and expanding subtrees lets the user navigate the program execution step by step. Clicking on raised rectangles one after the other lets the user explore the execution in increasing detail. But without a clear destination, the user is unlikely to uncover specific behavior through navigation alone.

Searching and filtering are invaluable in that respect, but we also provide visual guidance through several alternative renderings (or charts) for collapsed subtrees. For exam ple, a subtree may appearasa class legend (Figure 11) showing the kinds of objects in that subtree. When this is insufficient, the user can choose a more detailed chart showing a class communication graph (Figure 12). Notonly can he tell that a particular class of objects participate in a given subtree, but he also can see the classes with which they interact in the immediate context.

O ther chartmetaphors are possible, of course. A meter showing accumulative CPU time, an interor intra-class call matrix [9], and a histogram of instances [9] are shown schematically in Figure 13.

These reveal subtree inform ation with varying emphases and levels of detail. The pattern view accommodates charts of any size without significant rearrangement: conventional subtrees need only be displaced downwards.

## 4 Generalization

So far we have described some of the interactive features of the view that let the user expand, elide, and extract execution information. These are all visual manipulation techniques; the view supports nonvisual techniques as well. The view can be searched in several ways, and it can detect, generalize, and save recurring execution patterns. These capabilities work synergistically with the visual manipulation and navigation techniques, giving the user powerful tools for understanding program execution.

### 4.1 Why bother detecting recurrences?

We've already seen one kind of recurrence in Section 3.4, namely iteration. Execution patterns of iterative behavior rarely justify the space they consume. Combining them produces an aller, easier to understand views—just compare Figures 5 and 6. In most cases, Figure 5 provides little information of consequence over Figure 6, yet Figure 6 is considerably easier to assimilate. Not only does it reduce clutter; itm akes the iteration explicit even as it highlights the recurring computation.

The execution patterns that result from iteration and recursion aren'tm erely sim ilar; they are

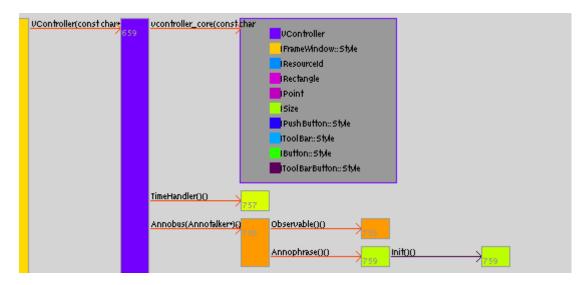


Figure 11: Class legends

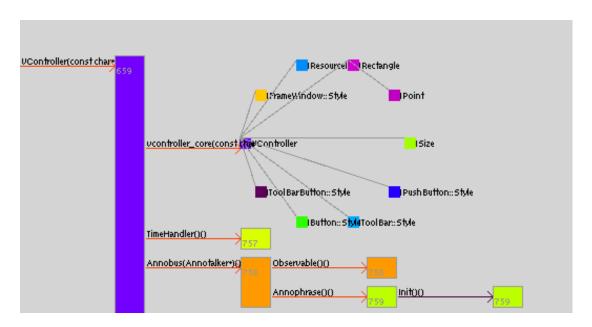


Figure 12:Class com munication graph

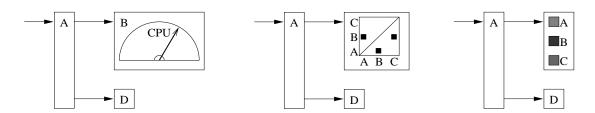


Figure 13:0 ther chartm etaphors

grouped together. Iterative execution patterns stack vertically, while recursive execution patterns chain horizontally. But recurrence isn't lim ited to iteration and recursion. Widely separated patterns may be similar or identical as well (due to parameterized types, for example). The more widely separated they are, them ore likely a user is to miss the recurrence.

A utom atic pattern detection is particularly importanthere. The system can attract our attention to far-flung yetsim ilarpatterns. Looking closely at the zoom ed-outview of Figure 8, for example, you might notice recurring splotches of coloratdistant intervals. A bird's-eye view makes it possible to see these distant recurrences, but it doesn't make doing so easy or accurate.

When you see an interesting pattern, you 'll probably want to find similar occurrences. Conversely, if you've seen a pattern and never want to see it again (because you're already familiar with it, or it does not concern you), then you'd want to hide its recurrences. The techniques we've described for filtered expansion (Section 3.3) filter per class or permethod. Filtering per pattern requires automatic recognition of pattern recurrences.

## 42 Why generalize?

M erely spotting recurrences is of limited use unless patterns are also generalized—that is, unless we can identify inexact recurrences among patterns. Searching for recurrences without generalization is like searching textwithout regular expressions: every match must be exact. We found that our first prototype, which did not generalize, didn't always combine patterns as we thought it should, because it didn't recognize a repetition of similar but not identical patterns. Often the differences among such patterns are immaterial; they are the same for the program mer's intents and purposes.

In fact, relatively few execution patterns in a typical program will be absolutely identical-involving the same objects, messages, and message order. Hence strictmenthing is too restrictive. Consider again the pattern shown in Figure 5, in which the sequencemore, cur, update, and next appears to recur six times. These recurrences are not really identical, however, because each repetition involves a different receiver of the update message. Without generalization, these patterns would not be identified as recurrent (as in Figure 6).

Butgeneralization prom isesm one than justa better job of recognizing loops. Measuring the average perform ance of just one execution pattern, for exam ple, is probably meaningless for optimization purposes. Not much time or space will be saved by optimizing one pattern. But the opportunities for improving performance expand considerably the more patterns recur. Generalization can identify potentially numerous recurrences that would otherwise be missed.

## 43 When are patterns the same?

To make generalization effective, we studied the situations in which two or more patterns might be deemed equivalent. That led to algorithms that could be tuned for different pattern matching criteria. As a result, the user can modulate generalization of patterns into higher-order patterns, thereby controlling the amount of information displayed. Seemingly complicated interactions often reduce to a handful of higher-order patterns, saving the user the trouble of examining multiple, trivially differentiated interactions.

The follow ing are the generalization criteria we found to matter most to program mers: identity, class identity, message structure, depth-limiting, repetition, polymorphism, associativity, and commutativity. A user can choose one or more criteria with which to combine patterns automatically.

#### 431 Identity

The sim plestdefinition of sim ilarity considers two execution patterns identical if their topologies are identical and if there is a perfect m atch for every object and every message in both patterns. This is easy to check, but it's too strict in general—few patterns match under this definition. Matching on identity is most useful for finding exact matchesthat is, when the program mer knows precisely what to look for.

#### 432 Class Identity

Often, the sam e pattern of messages affects different sets of objects. If the same message gets sent to the same class of object, the program merwill interpret the sends as identical behavior. Therefore it's useful to consider messages sent to different objects as part of the same pattern if the rest of the pattern is the same and the receiving objects are of the same class. We emphasize the similarity of

such patterns visually by associating colors with class types. Two patterns match with respect to class identity if they have the same graph structure and identical coloring.

Consider the execution pattern associated with instantiating an object in Smalltalk [11] by executing Rectangular new. The pattern involves the class object Rectangular (of the metaclass Class) and a new ly allocated object of class Rectangular. Distinct instantiations do not follow identical patterns because each involves a particular instance of Rectangular. But to the programem er, these instantiations are almost always identical conceptually. Thus it's useful to characterize allof them with a more general pattern, which we might call the "Rectangular instantiation" pattern.

M ore precisely, if two patterns have isom orphic invocation graphs (with messages labeling the edges and types labeling the nodes), then they can be considered instances of the same pattern. Most program mers will consider such patterns similar and willprefer to see the generalized pattern. Nevertheless, when we present a generalized pattern, we still allow the user to query the actual object identity by brow sing through different instances of the pattern in the trace.

This technique is applied, for example, in depicting repetitions. If you look at Figure 6 closely you will notice that the green and black objects no longer show an object identifier. That's because they represent multiple instances, as shown in Figure 5.

#### 433 Message Structure

In a weakly typed language like Smalltalk, the type of the message receiver is not always known a priori. While the Rectangular instantiation pattern matches instantiation of different Rectangular objects, instantiation of a Rectangular and a Circle would not be considered the same, even though both invoke the same method new defined in the metaclass Class. We can generalize the matching by considering two patterns the same if their message structure is identical and there is a nontrivial color substitution from one's coloring to the other. This resembles the notion of alternations in regular expressions.

## 434 Depth-Limiting

Encapsulation is a key concept in object-oriented program m ing. It lets a program m er think in term s of w hat an object does (i.e., its interface), not how it does it (its in plem entation). Encapsulation w orks at multiple levels, especially in large, layered system s. O bjects in one layer com municate w ith objects in the layer below them exclusively through their interfaces.

Such layering provides another criterion for controlling matches. If a program is designed with good layering and encapsulation, patterns involving low er-level objects should be independent of those involving higher-level objects. Thus we can apply different matching criteria to patterns on the basis of message depth. A simple approach ignores matches to patterns beyond a given depth in the tree.

Continuing our Sm alltalk example, the Rectangular instantiation pattern begins with a message new received by the class object Rectangular. This invokes them ethod new defined in Class. In response, Rectangular sends itselfnew: to allocate a new object, and then it forwards them essage new to the new object. That in turn invokes Rectangular > new. Object creation may therefore span a deep execution graph during its initialization. By limiting the depth of the instantiation pattern, we can match instantiations comprising varied initialization patterns.

## 435 Repetition

Suppose one invocation of a loop executes it 1000 times, and a different invocation executes it only 999 times. (This behavior might be characteristic of a search algorithm that looks for a word somewhere in a long string, for example.) Most program mers would consider the overall behavior of the two invocations to be pretty similar. The loop would appear as a repeated structure along the vertical dimension (characteristic of iteration) in the execution pattern view. An example of repetition in the horizontal direction (recursion) could be an instantiation of object lists, where each object instantiates its neighbor. A list of six objects should reflect the same instantiation pattern as would a list of eight objects.

The matching algorithm can ignore the number of repetitions in a low er-level pattern when matching at a higher level. Hence if two patterns contain invocations of the search algorithm on dif-

<sup>&</sup>lt;sup>1</sup>An example of a "trivial" color substitution would be one that maps all colors to black, which has the same effect as ignoring color altogether.

ferentwords, they may be considered instances of a more general pattern. This resembles the notion of repetition in regular expressions.

#### 43.6 Polymorphism

Consider a program that draws a series of shapes, such as rectangles and circles. This can show up in an execution pattern as an iteration of polymorphic draw messages to different Shape objects. Rather than showing the classes of these objects, we match all these subclasses according to their common base class Shape.

Returning once m ore to the Sm alltalk exam ple, Circle new and Rectangular new generate polym orphic instantiation patterns provided both are subclasses of Shape. Of course, m atching two classes when their base class is the same does not imply that we will match two different method names. This suggests a more general definition of polym orphicm atching that will match methods with overloaded names as well.

#### 43.7 Associativity

W hat if the colors of two patterns m atch but their structures don't? Structural equivalence can take into account properties such as associativity of objects or m ethods.

If a m athem atical function  $\varphi$  is associative, then nested invocations can be flattened by effectively rem oving the inner sets of braces:  $\varphi(x,\varphi(y,z))=\varphi(\varphi(x,y),z)=\varphi(x,y,z)$ . M athem atica [23], for exam ple, uses this property to get functions into a standard form before m atching them to patterns.

Nested lists are a good illustration: Flatten [ $\{C, \{D\}\}, E\}$ ] flattens out the sublists at all levels, resulting in  $\{C, D, E\}$ . In matching execution patterns, we can treat an "associative" class A in a manner analogous to the braces of the nested list above. Consider the leftmost pattern in Figure 14. Removing the inner calls of A would allow it to match the rightmost pattern in that figure, wherein the first call to A calls C, D, and E directly. This would let you ignore all private method calls in class A, for example, when A's in plementation of a public service is of no concern. A ssociativity thus helps suppress successive self-invocations.

When execution patterns are matched with respect to associativity of classes, method labels are ignored. Conversely, we can choose to ignore object type and/or identity and match patterns on associativity of methods exclusively. If a

m ethod is associative, then a sequence of recursive invocations— even on different objects— m atchesa single call.

Recall (Section 3.6) that the execution pattern view let the user choose how a class is displayed based on its attributes. A natural rendering (and the system 'sdefault) for expressing associativity is flattening, as shown in Figure 14.N ote how C, D, and E are aligned vertically as a result of flattening A. The depth of the tree no longer represents physical stack depth but rather a logical one—one that ignores self-invocations. This example also demonstrates how we carefully match the elision technique—flattening in this case—to the need, namely association.

Figure 14 also reveals a difficulty. Had C and D (or D and E , or both) been the same class, flattening A would make them match based on repetition.

#### 438 Commutativity

Com m utativity m eans that a pattern wherein A is called first and then B m atches a pattern wherein B is called first and then A. To avoid comparing every combination of calls whenever a commutative m atch is desired, objects acquire their commutative characteristics from their context: two subtrees m atch on commutativity if they involve instances of the same set of classes. This limits the possiblem atchings to complete subtrees. Thus if our visualization displays subtrees as orderless lists of classes as depicted in Figure 11, then two subtrees that m atch based on commutativity will be rendered identically.

## 5 Im plem enting G eneralization

A tracing tool collects "m ethod enter" and "m ethod leave" events from the target program as itruns. The resulting trace m ay drive visualizations in either real-time or post-m or tem.

From the information in the trace, the view builds a tree structure representing the sequence of messages and the objects that receive them. Every node in the tree structure may represent an execution pattern. To recognize, classify, and generalize execution patterns, the view assigns a hash value to every node in the tree structure. A set of tree nodes reflecting a generalized pattern will have identical hash values.

A recursive hash function com putes a hash value for each subtree in a single pass over the structure.

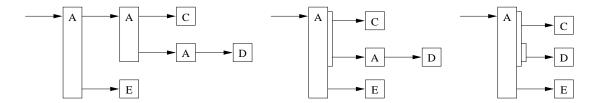


Figure 14: Using flattening to visualize association

The hash value of a given subtree is a function of its children is hash values and one or more values in the subtree is root. Which values are used depends on the matching criterion. If we want to match on method names, the values include the method and class names. If we want to match on class names only, then the method name is omitted.

Sim ilarly, if we want to match on a particular object, then that object's ID must be included. If we're interested in knowing not the total number of method calls but just them ethods that got called, we can ignore nodes representing redundant calls during recursion. If we don't care about the order of calls, we can sort the nodes first.

Using a Universal, class of hash functions [6] ensures that computing the hash value recursively will produce a good distribution. The chance of two different patterns producing the same hash value is no more than the depth of the tree divided by the maximum hash value. Most hash values used in this computation are too large to use as an index into a table—typically around  $2^{32}$ . So each hash value is masked down to a reasonable number of bits for an index.

The masked hash value is registered in a pattern dictionary, which contains all the hash values. Each entry in this dictionary corresponds to a (perhaps trivially) generalized execution pattern. When the system encounters a hash value that salready registered, it increments one ormore fields associated with the dictionary entry for that value—for example, a frequency field recording the total number of incidences of this pattern. Other cumulative information such as CPU time may be stored to help further characterize and differentiate the program 's execution patterns.

For good perform ance, our prototype checks only that hash values are identical to establish equivalence. This can create false positives, but the distribution is good enough to make this rare, and perform ance is considerably in proved over more detailed comparisons.

Som etim es it is im practical to show method names in graphs due to a lack of space, an unworkably large number of methods, or both. We use two tricks to mitigate this problem: zooming and "flyover." It's possible to zoom the view without scaling the fonts, providing more room for method names at higher magnifications. We also identify the method (and/orobject) directly under the mouse cursor in separate status panels. When space is at a premium, we om it method names and other labeling and let the user examine methods selectively by "flying over" them with the mouse.

## 6 Related Work

Our work is a confluence of two research areas: pattern m atching and program visualization. Both have long histories, but they have been independent until recently.

Tree and string m atching is one of the most thoroughly documented areas of research in computer science [13,21,5,19,7,4,1]. The algorithms are many, they are well-understood, and they have been applied widely—in interpreters for nonprocedural languages; optimizing compilers; algebraic computation; sorting, searching, and differencing facilities of all types; and automatic theorem proving. We have tailored a few of these algorithms to the needs of our domain. Pattern extraction has in factbeen applied to program sbefore, but almost exclusively as a static analysis (as in SCRUPLE [20], for example).

We can classify most object-oriented visualization systems into two categories: macroscopic and microscopic. Macroscopic systems collect and present cumulative execution information, while microscopic systems help you understand the sequence of message sends between objects. Our earlier visualization work is characteristic of the macroscopic type [9, 10]; the microscopic variety is more common [17, 12, 8, 2].

This dichotom y reflects how hard it is to present execution information effectively. Macroscopic system s condense execution inform ation to make it m ore m anageable, m uch like traditional profiling tools do- and they discard much information in the process. Our original visualization system offered severalview sofm essage sends per class, the overall activity of objects, and resource (CPU and m em ory) consum ption. These cum u lative view s could be navigated to uncoverm ore detail- for exam ple, totalm essage traffic at the m ethod (rather than class) level. Still, this didn't provide much insight into how the program accomplishes a particular task, such as initialization or screen update. We were discarding execution information at the object level in the name of scalability.

M icroscopic system s show the sequence ofm essages between objects, potentially yielding deep insights about small sections of the program. But m icroscopic systems have the opposite problem from theirm acroscopic counterparts: toom uch detail limits scalability. Showing individual objects sending messages to each other (typically using a nodes-and-arcsmotif) quickly gets unworkable as the number of objects increases. Not only is there a multitude of indistinct objects, but their communication is equally obscure. The messages and messaging paths are lost in a jumble of lines and bubbles. Any benefits of visualization are quickly lost.

Only recently have people tried to narrow the gulf between these extremes. Program Explorer [18] is a representative example; it im plem ents Jacobson's interaction diagram snearly verbatim to visualize object interaction. The system em ploys several filtering techniques to help manage large num bers of objects. But since every object in the visualization occupies a column from the top to the bottom of the view, scaling remains a problem . M oreover, the interaction diagram becom es unw ieldy w hen objects com m unicate w ith others that were created much later, again because rails appear in the order of object creation. Hence when there is communication between objects created at distant intervals, the screen fills with long horizontallines spanning potentially many screenfuls. This problem is not easily solved. Rearranging the vertical lines that denote objects is highly disruptive; grouping the objects reintroduces the shortcom ings of macroscopic systems.

Our execution pattern view offers m any of the advantages of both m icro- and m acroscopic ap-

proaches. It can provide cum ulative orglobal execution inform ation by collecting and generalizing patterns throughout the execution trace. It also lets the user inspect the program at any level of detail with its navigation (elision and expansion) features. It is not perfect—accurate pattern generalization being the most challenging aspect—but it does offer some new choices on the micro-tomacroscopic visualization spectrum.

## 7 Conclusion

Execution patterns enhance object-oriented visualization technology in three ways. First, they offer an intuitive and scalable metaphor for object communication. Our execution pattern view 's structure is based on a natural notation, one that captures object interaction clearly and lends itself to interactive manipulation. One can easily navigate the execution trace to survey the interactions of objects, classes, and methods. The pattern view differs from previous views in that it scales smoothly to make even lengthy interactions of objects intelligible. Elision and expansion mechanisms make the journey from the macroscopic to the microscopic easy for users.

The second enhancement comes from generalizing similar execution patterns. Generalization letsus abstractaway redundant behaviorautomatically, even when the redundancy is imprecise or non-periodic.

Third, execution patterns give us a foothold for characterizing system complexity. While the definitive complexity metric for execution patterns is far from obvious, we have found even simpleminded metrics (e.g., pattern redundancy) useful for pinpointing not just any program hotspots but those that are most likely to have a simple remedy.

We have implemented the execution pattern view and integrated it into 0 vation [9,10], our research prototype for object-oriented program visualization. The views in this paper were taken from traces of 0 vation itself. The system can visualize any C++ or Java program using traces generated from the VisualAge development environment [14]. O vation also supports visualization of Smalltalk programs. To generate Smalltalk traces, we added instrumentation to the Little Smalltalk [3] and VisualAge Smalltalk [15] environments.

We have experimented with traces from midsize programs such as 0 vation itself and truly large system s such as Taligent. We found that the view and mechanisms described here were helpful in uncovering unexpected behavior, in understanding unfamiliar code, and in improving performance.

Currently we are making the pattern matching facility more flexible. We are augmenting the matching criteria with a pattern equivalent of regular expressions to provide a semiformal way to express similarity. We are also exploring the potential of visual grammars, especially their synergy with the current visualizations. Finally, we plan to report qualitative results of execution pattern visualization as we learn through user feedback and controlled experiments.

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