



the ray kernel
 soot · eclipse · svn
 weka · java · symex
 instrumentation
 readability · docinf · ast
 frequency · complexity
 WRG



Snippet Sniper
 Helix



[ARRESTED COMPUTING]



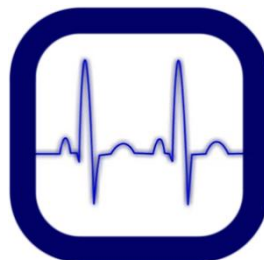
Computer Science
 at the UNIVERSITY of VIRGINIA

A Human-Centric Approach to Program Understanding

“The real question is not whether machines think,
 but whether men do.” -- B. F. Skinner



Readability



Runtime Behavior

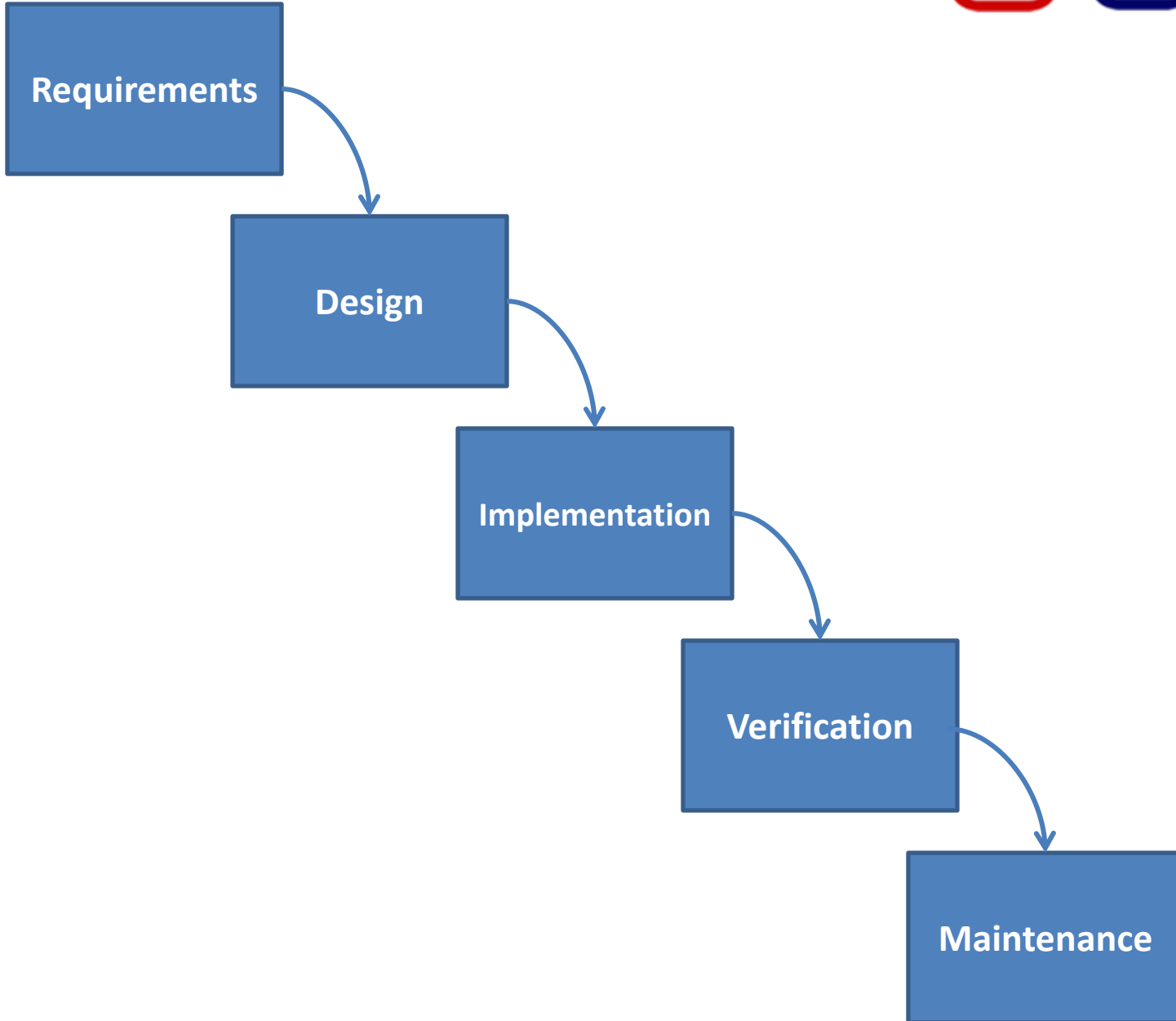


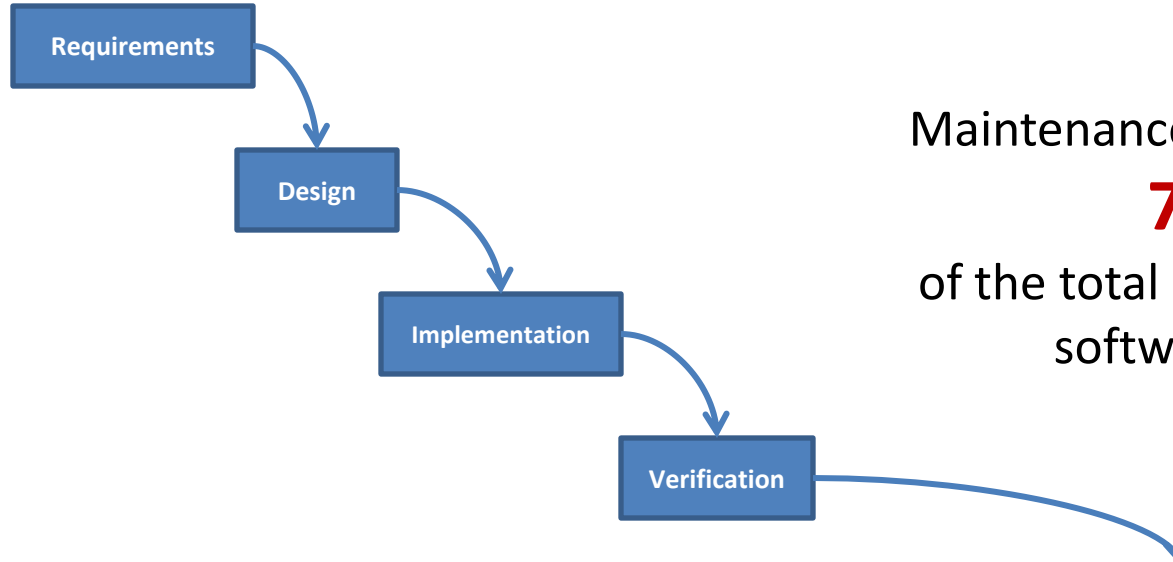
Documentation

Ray Buse - PhD Proposal

University of Virginia, Department of Computer Science

1.20.2010



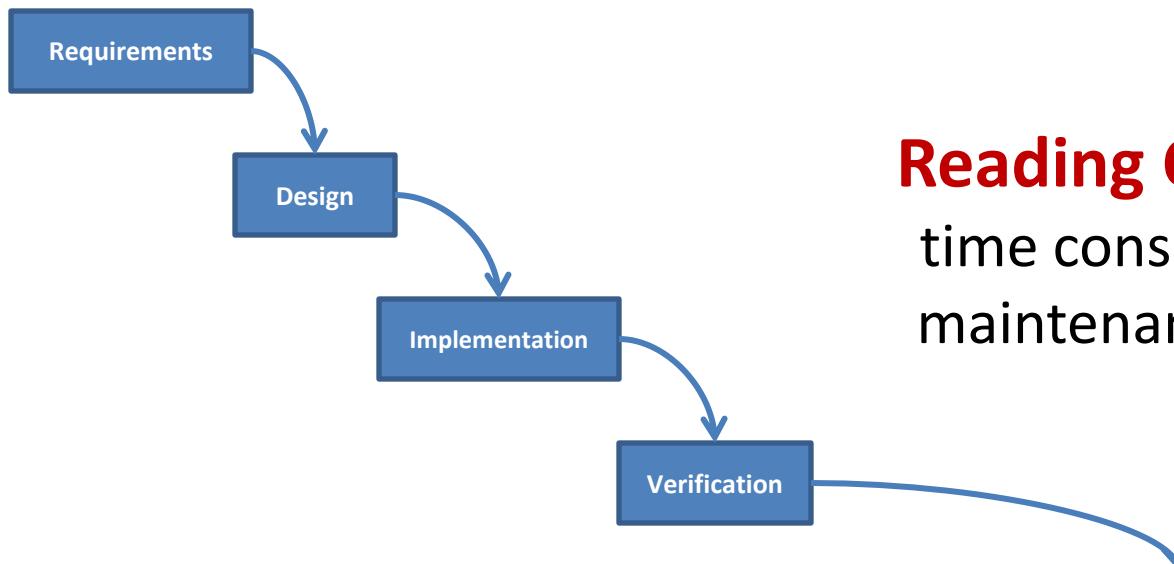


Maintenance accounts for about
70-90%
of the total lifecycle budget of a
software project.^{1,2}

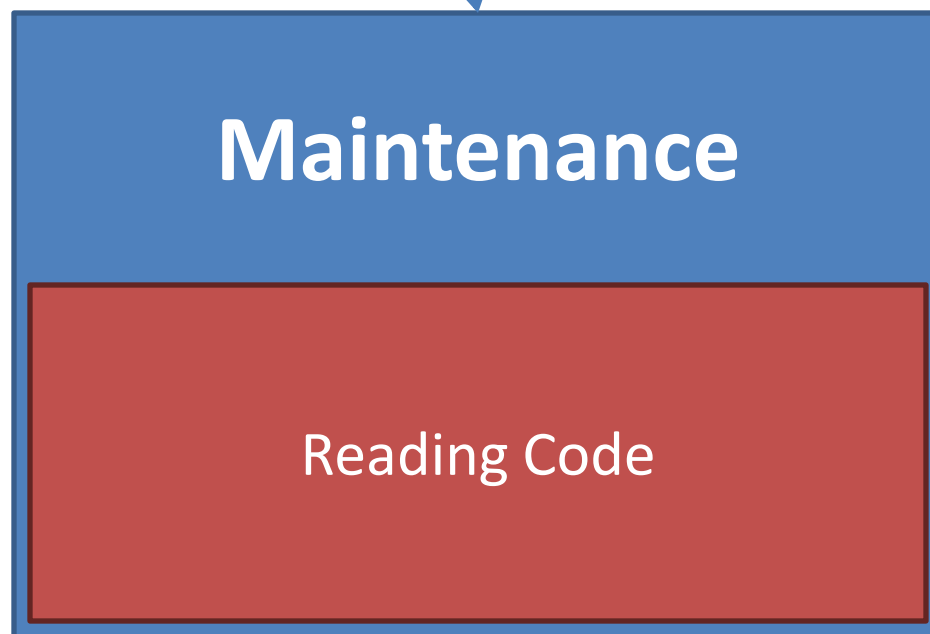
1. T. M. Pigoski. *Practical Software Maintenance: Best Practices for Managing Your Software Investment*. John Wiley & Sons, Inc., 1996.

2. R. C. Seacord, D. Plakosh, and G. A. Lewis. *Modernizing Legacy Systems: Software Technologies, Engineering Process and Business Practices*. Addison-Wesley Longman Publishing Co., Inc., Boston, MA, USA, 2003.

Maintenance



Reading Code is the most time consuming part of all maintenance activities.^{3,4,5}



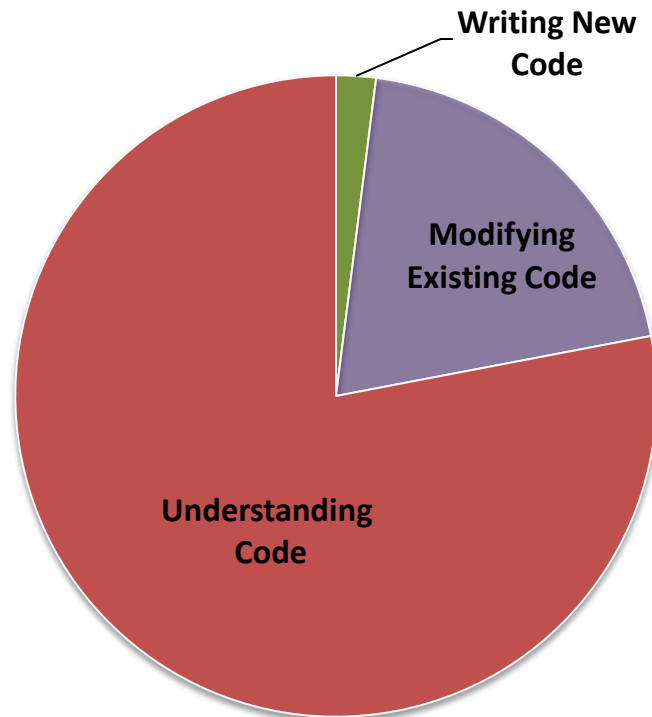
3. L. E. Deimel Jr. *The uses of program reading*. SIGCSE Bull., 17(2):5-14, 1985.

4. R. Glass. *Facts and Fallacies of Software Engineering*. Addison-Wesley, 2003.

5. S. Rugaber. *The use of domain knowledge in program understanding*. Ann. Softw. Eng.,(1-4):143-192, 2000.



“Understanding code is by far the activity at which professional developers spend most of their time.” ⁶



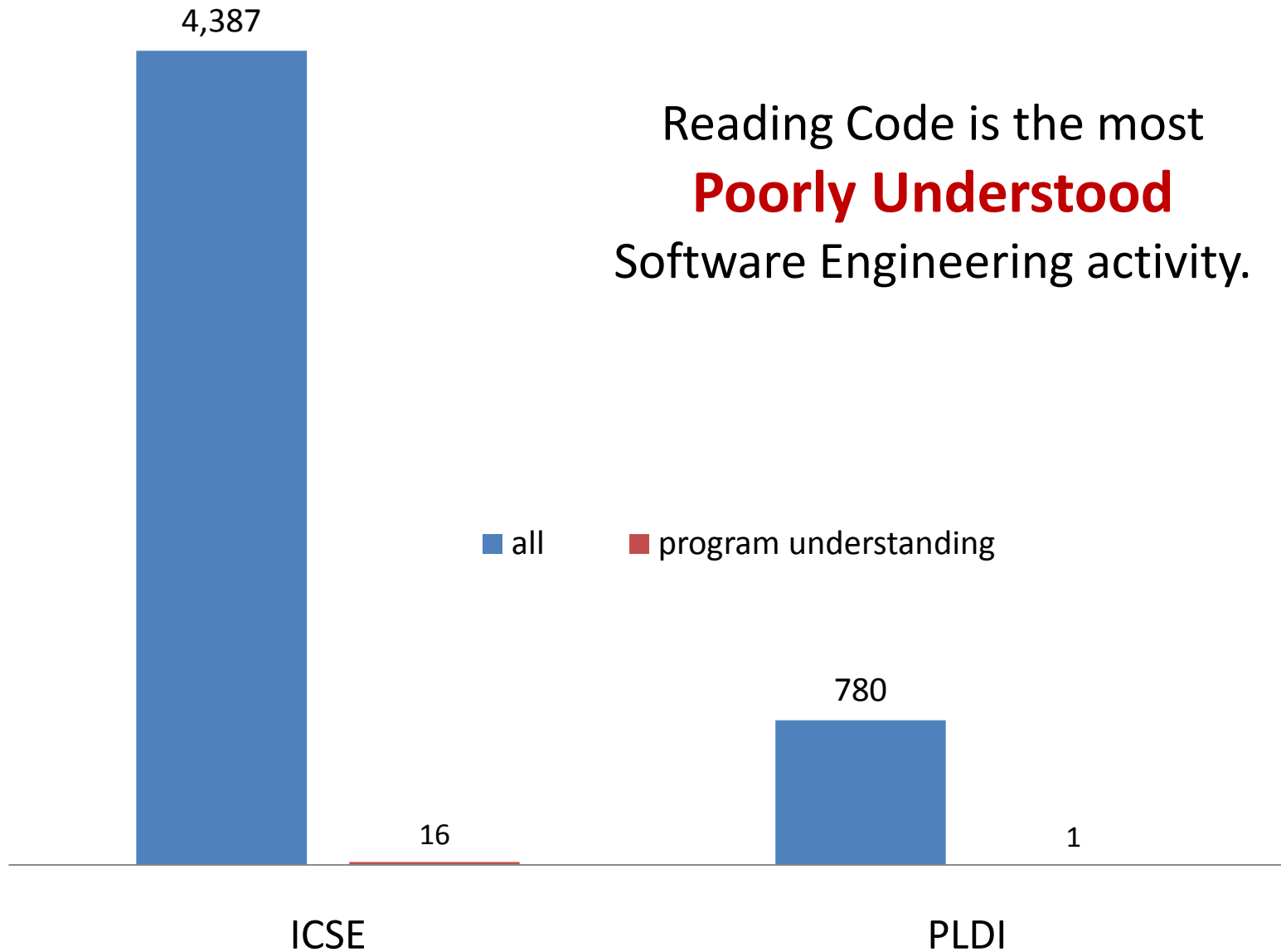
6. Peter Hallam. *What Do Programmers Really Do Anyway?* Microsoft Developer Network (MSDN) – C# Compiler. Jan 2006.



Reading Code is the most
Poorly Understood
Software Engineering activity.^{7,8}

7. D. Parnas. *Software aging*. In *Software Fundamentals*. Addison-Wesley, 2001.

8. D. Zokaities. *Writing understandable code*. In *Software Development*, pages 48-49, jan 2002.





Understanding is difficult to...

Model



- Based on a complex combination of factors

Evaluate

- Lack of established metrics/baselines
- User studies are unattractive



Two Key Insights

-  **Machine Learning** allows us to combine many semantically **shallow features** of code to gain new **deep insights**.
-  **PL Techniques** can be adapted to generate documentation artifacts that are **directly comparable** to human created ones.



Thesis

We can combine insights from **Machine Learning** and **Programming Languages** to

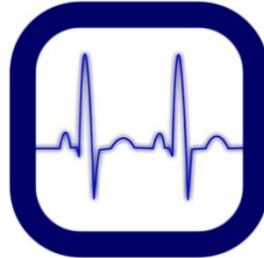
- **Model** aspects of code understanding *accurately* and
- **Generate** output that compares favorably with human documentation.



Proposal: Three Dimensions of Understanding



Readability



Runtime Behavior



Documentation



Proposal: Three Dimensions of Understanding



Readability

Textual characteristics that make code **understandable**.



Runtime Behavior

Structural characteristics that help developers **understand** what a program is expected to do.



Documentation

Non-code text that helps developers **understand** a program.



Research Projects

Metrics for:

- Code Readability
- Path Execution Frequency

Algorithms for Documentation of:

- Exceptions
- Code Changes
- APIs



Broader Impact

New algorithms and metrics to support:

- Software Development and Composition
 - Metrics for Software Quality Assurance
 - Automatic Documentation
- Software Analysis
 - Runtime Behavior model for optimizing compilers
 - Metrics for targeting analyses, prioritizing output, and evaluating research



The rest of this proposal

- A review of each proposed contribution
 - Technical Merit
 - Evaluation Strategy
 - Related Work
- Research timeline and other bookkeeping
- Concluding Remarks



Metrics for:

- **Code Readability**  ISSTA '08  TSE '10
- **Path Execution Frequency**  ICSE '09

Algorithms for Documentation of:






- **Exceptions**  ISSTA '08
 - **Code Changes** 
 - **APIs** 
-  Published
 In Progress



Metrics for:

- **Code Readability**  ISSTA '08  TSE '10
- Path Execution Frequency  ICSE '09

Algorithms for Documentation of:

- Exceptions  ISSTA '08
 - Code Changes 
 - APIs 
-  Published
 In Progress



Readability

Model human judgments about code readability

Create a readability metric

Key: Use textual features to *approximate* human judgments

```
/**
 * Extend this Execution path by one level.
 *
 * @throws IllegalStateException If the move path invalid..
 */
private List<ExecutionPath> extend (ExecutionPath ep)
{
    paths = new LinkedList<ExecutionPath>();

    Unit last = ep.getLast();

    List<Unit> succs = graph.getSuccsOf(last);

    //this is the end of the path
    if (succs.isEmpty())
    {
        ep.setComplete(true);
        paths.add(ep);
        return paths;
    }

    if (succs.size() == 1)
    {
        Unit s = succs.get(0);
        if (ep.contains(s))
        {
            //do nothing
        }
        else
        {
            ep.addLast(s);

            if (graph.getTails().contains(s))
            {
                ep.setComplete(true);
            }
        }
    }
}
```



Hypothesis

With a simple set of **textual features**, we can derive from a set of **human judgments** an **accurate model** of readability for code.

Success depends on

- Gathering human judgments
- Choosing predictive textual features




Data Gathering

- We asked 120 students at UVa to rate the readability of a set of snippets...

```
-----  
/**  
 * Computes factorial with recursion  
 */  
public int factorial( int integer )  
{  
    if( integer < 1 )  
        return 0;  
  
    if( integer == 1)  
        return 1;  
  
    return integer * factorial( integer - 1 );  
}  
-----
```

Snippet Pack demo: 2 of 4

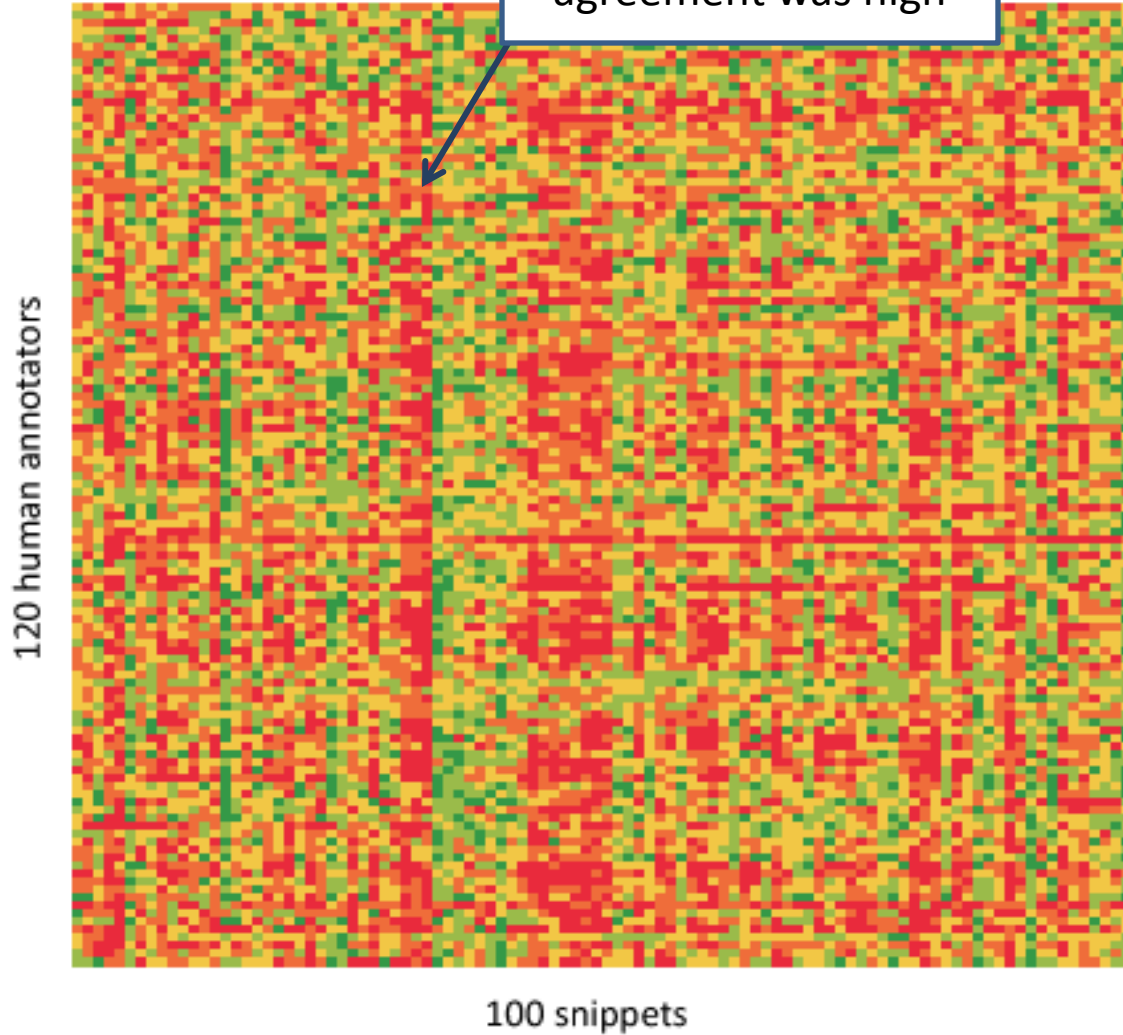
1 2 3 4 5

 ← → 



Data Set

Vertical bands indicate snippets were agreement was high





Choosing predictive textual features

We choose **local** code features

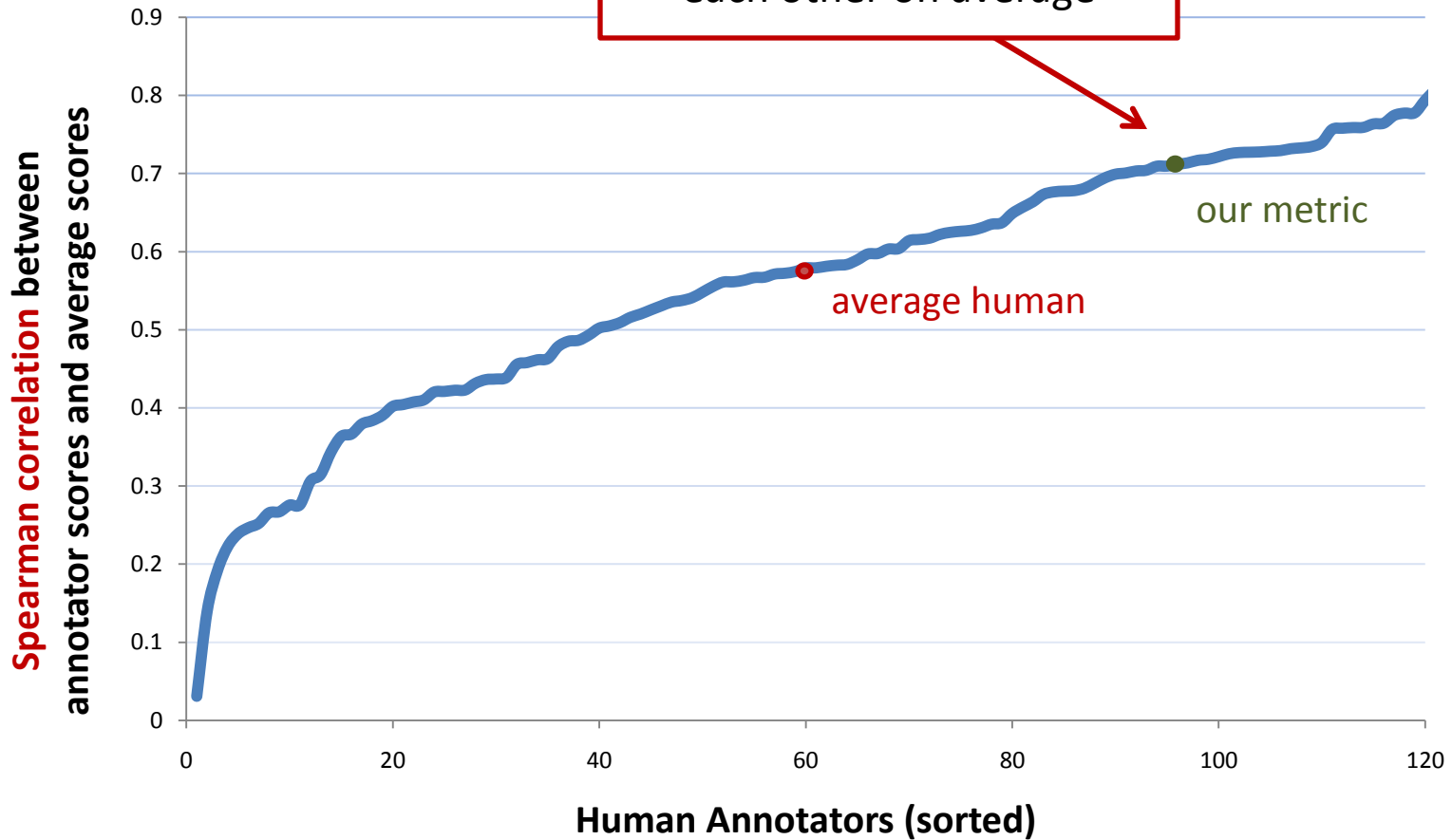
- Line length
- Length of identifier names
- Comment density
- Blank lines
- Presence of numbers
- [and 20 others]

Modeled with a Bayesian Classifier



Model Performance

Model agrees with humans as much as they agree with each other on average





Related Work

- Readability metrics for natural languages
 - Very popular, DOD standards etc
- In the software domain
 - Complexity metrics (often used, but utility is questionable)



Conclusions






- We can automatically judge readability about as well as the *average* human can
- This notion of readability shows significant correlation with:
 - Code churn
 - A bug finder
 - Program maturity

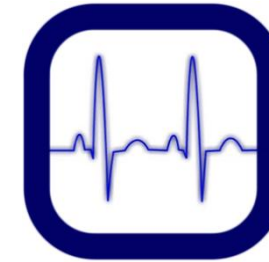


Metrics for:

- Code Readability  ISSTA '08  TSE '10
- **Path Execution Frequency**  ICSE '09

Algorithms for Documentation of:

- Exceptions  ISSTA '08
- Code Changes 
- APIs   Published
 In Progress



Runtime Behavior

Model path execution frequency statically

Key: Use path surface features to uncover developer expectations

```
/**
 * Extend this Execution path by one level.
 *
 * @throws IllegalStateException If the move path invalid..
 */
private List<ExecutionPath> extend (ExecutionPath ep)
{
    paths = new LinkedList<ExecutionPath>();
    Unit last = ep.getLast();

    List<Unit> succs = graph.getSuccsOf(last);



    //this is the end of the path
    if (succs.isEmpty())
    {
        ep.setComplete(true);
        paths.add(ep);
        return paths;
    }

    if (succs.size() == 1)
    {
        Unit s = succs.get(0);
        if (ep.contains(s))
        {
            //do nothing
        }
        else
        {
            ep.addLast(s);

            if (graph.getTails().contains(s))
            {
                ep.setComplete(true);
            }
        }
    }
}
```



Key Idea

-  Developers often have **expectations** about common and uncommon cases in programs
-  The **structure** of code they write can sometimes reveal these expectations



Intuition

```
public V put(K key , V value)
{
    if ( value == null )
        throw new Exception();

    if ( count >= threshold )
        rehash();

    index = key.hashCode() % length;

    table[index] = new Entry(key, value);
    count++;

    return value;
}
```

Exception

Invocation that changes a lot of the object state

Some computation



Hypothesis

We can *accurately* predict the runtime frequency of program *paths* by analyzing their static *surface features*

Goal:

- Know what programs are *likely* to do without having to run them (produce a *static profile*)



Applications for Static Profiles

Indicative (dynamic) profiles are often unavailable

Profile information can improve many analyses

- Profile guided optimization
- Complexity/Runtime estimation
- Anomaly detection
- Significance of difference between program versions
- Prioritizing output from other static analyses

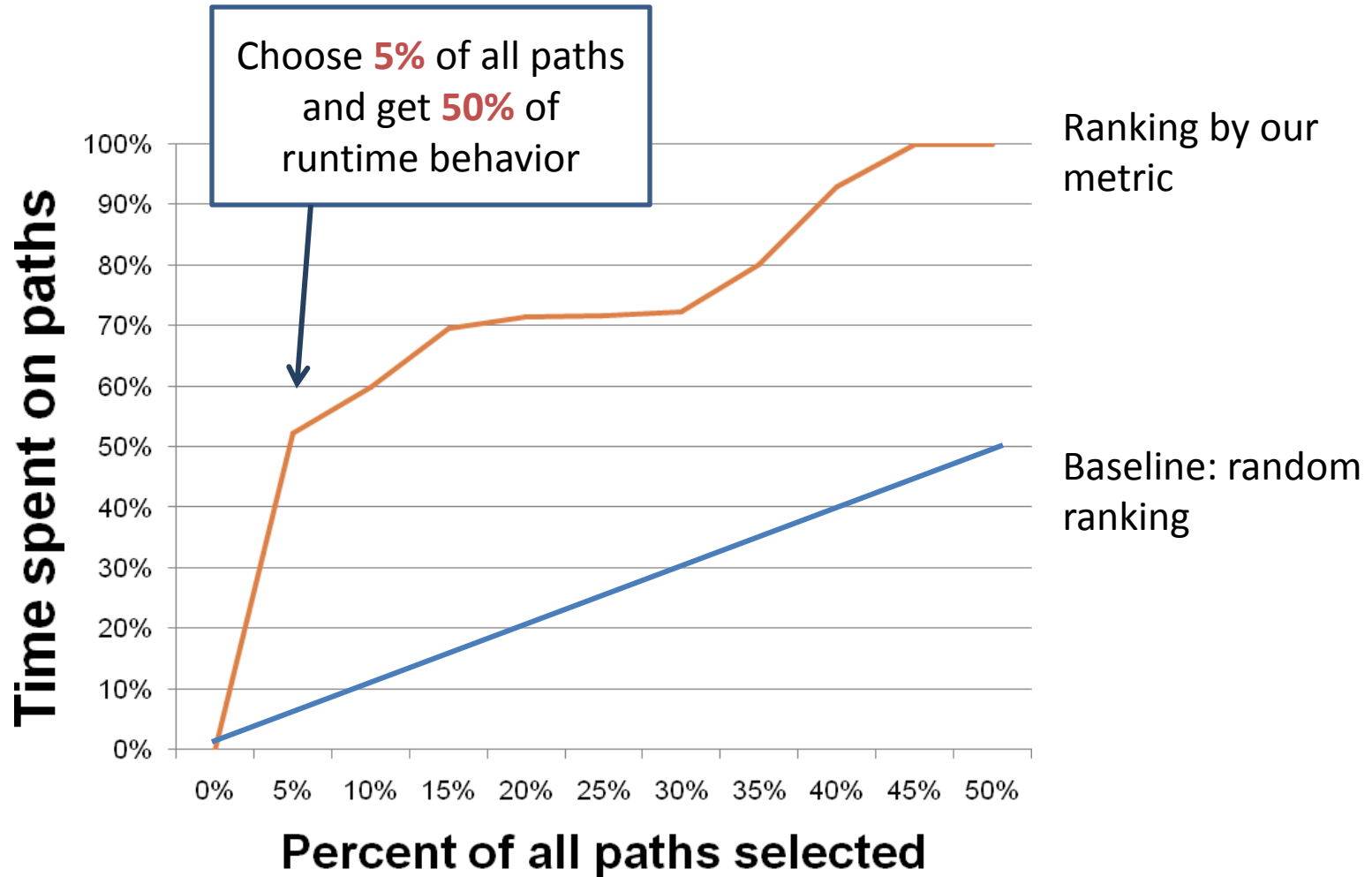


Approach

- **Model** path with a set of features that may correlate with runtime **path** frequency
- **Learn** from programs for which we have indicative workloads, we used a Logistic Regression
- **Predict** which **paths** are most or least likely in other programs

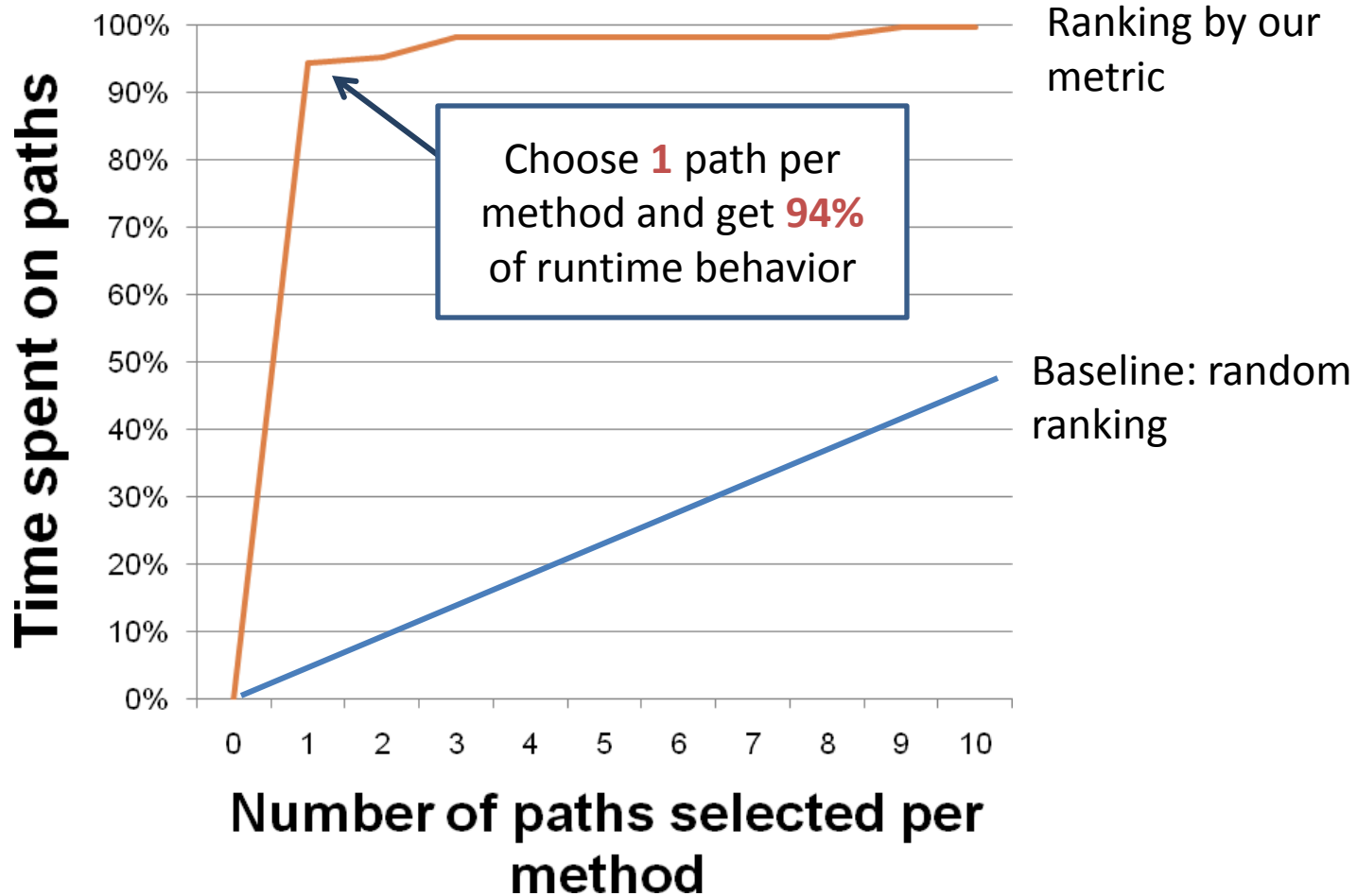


Evaluation





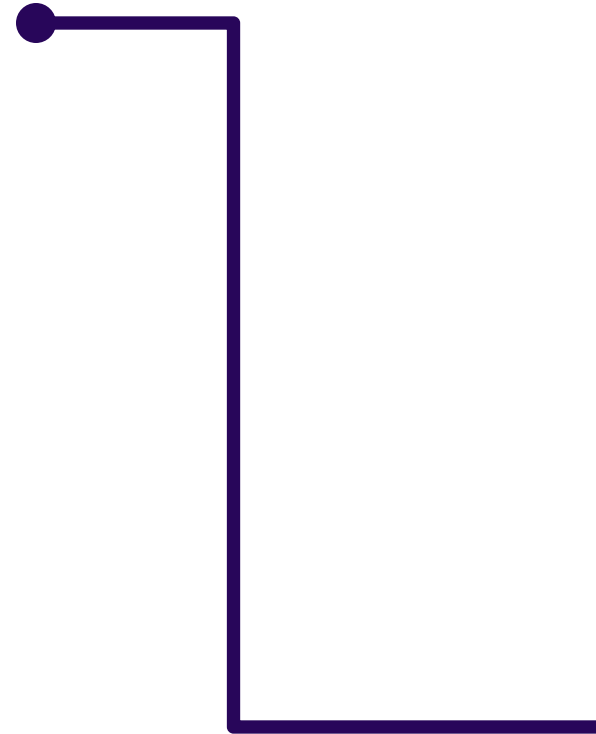
Evaluation





Related Work

- Static Branch Prediction [Ball & Larus '92]
 - For each branch, which direction is most likely
 - In a direct comparison, our tool is better





Conclusion

- A formal model that statically predicts relative dynamic path execution frequencies
- The promise of helping other program analyses and transformations



Metrics for:

- Code Readability  ISSTA '08  TSE '10
- Path Execution Frequency  ICSE '09

Algorithms for Documentation of:

- Exceptions  ISSTA '08
- Code Changes 
- APIs   Published
 In Progress



```
/**
 * Extend this Execution path by one level.
 */
/**
 * Extend this Execution path by one level.
 * @throws IllegalStateException If the move path invalid..
 */
private List<ExecutionPath> extend (ExecutionPath ep)
{
    paths = new LinkedList<ExecutionPath>();
    Unit last = ep.getLast();
    List<Unit> succs = graph.getSuccsOf(last);
    //this is the end of the path
    if (succs.isEmpty())
    {
        ep.setComplete(true);
        paths.add(ep);
        return paths;
    }
    if (succs.size() == 1)
    {
        Unit s = succs.get(0);
        if (ep.contains(s))
        {
            //do nothing
        }
        else
        {
            ep.addLast(s);
            if (graph.getTails().contains(s))
            {
                ep.setComplete(true);
            }
        }
    }
}
```



Documentation

Generate for:

- Exceptions
- APIs
- Version Changes

Key: Use symbolic execution and summarization heuristics to generate human-readable results.



Use

- For Internal Developers
 - Easier to **keep track** of what's going on
- For Maintenance and Testing
 - Easier to **read** old code.
- For External Developers
 - Easier to **integrate** off-the-shelf software libraries



Three Types of Documentation

- Exceptions
- Code Changes
- APIs



Documenting Exceptions

```
/**  
 * @throws Exception If the value is null  
 */  
public V put(K key , V value)  
{  
    if ( value == null )  
        throw new Exception();  
  
    if ( count >= threshold )  
        rehash();  
  
    index = key.hashCode() % length;  
  
    ...  
}
```

Best practice dictates that exceptions should be documented



Documenting Exceptions

```
/**
 * @throws Exception If the value is null
 */
public V put(K key , V value)
{
    if ( value == null )
        throw new Exception();

    if ( count >= threshold )
        rehash();

    index = key.hashCode() % ...
    ...
}
```

Best practice dictates that exceptions should be documented

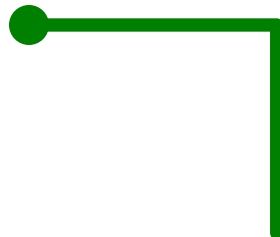
Does this method throw an exception?



Importance

Mishandling or Not handling can lead to:

- Security vulnerabilities
- May disclose sensitive implementation details
- Breaches of API encapsulation
- Any number of minor to serious system failures





Hypothesis

Mechanical documentation of exceptions can be *at least as good* as human on average.

- More *complete*
- More *accurate*

We extract paths to **throw** statements and use symbolic execution to *generate path predicates*



Examples

- Sometimes we do **better**:

```
Worse: id == null  
(Us) Better: id is null or id.equals("")
```

- Sometimes we do about the **same**:

```
Same: has an insufficient amount of gold.  
(Us) Same: getPriceForBuilding() > getOwner().getGold()
```

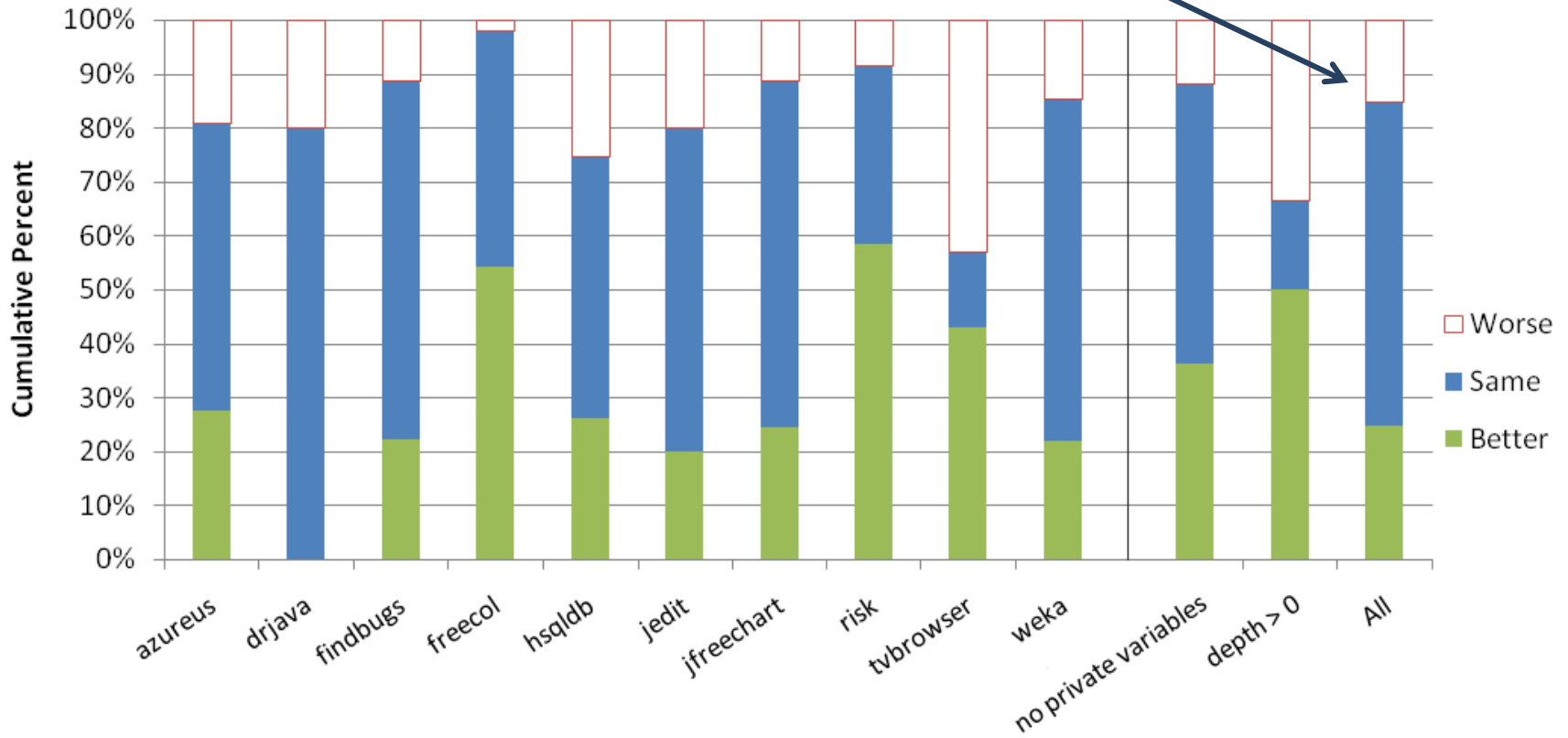
- Sometimes we do **worse**:

```
Better: the queue is empty  
(Us) Worse: private variable m_Head is null
```



Key Results

Our documentation is as good as human over **80%** of the time





Code Change Examples



mercurial



jfreechart rev 3405

(start): Changed from Date to long,
(end): Likewise,
(getStartMillis): New method,
(getEndMillis): Likewise,
(getStart): Returns new date instance,
(getEnd): Likewise.

Phex 3542

Minor change

Jabref rev 2917

Fixed NullPointerException when
downloading external file and file directory
is undefined.



Toby,
Going forward, **could you I ask you to be more descriptive in your commit messages?** Ideally you should state what you've changed and also why (unless it's obvious)... I know you're busy and this takes more time, but it will help anyone who looks through the log ...

<http://lists.macosforge.org/pipermail/macports-dev/2009-June/008881.html>

Subject: An appeal for more descriptive commit messages
I know there is a lot going on but **please can we be a bit more descriptive when committing changes.** Recent log messages have included:

"some cleanup"

"more external service work"

"Fixed a bug in wiring"

which are a lot less informative than others...

<http://osdir.com/ml/apache.webservice.tuscany.devel/2006-02/msg00227.html>

Sorry to be a pain in the neck about this, but could we **please use more descriptive commit messages?** I do try to read the commit emails, but since the vast majority of comments are "CAY-XYZ", I can't really tell what's going on unless I then look it up.

<http://osdir.com/ml/java.cayenne.devel/2006-10/msg00044.html>



Key Idea



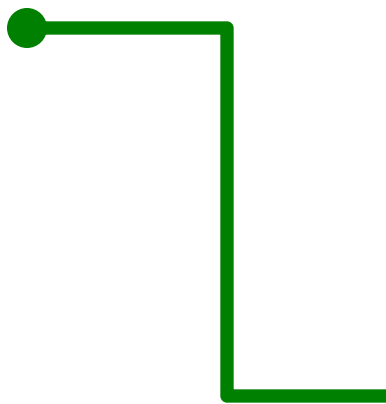
Generate Documentation that describes the **effect** of a change on the runtime behavior of a program

- What conditions are necessary to activate the change
- What the new behavior is



Algorithm

- Generate predicates for each statement
- Compare predicates across versions
- Summarize change and distill structured output

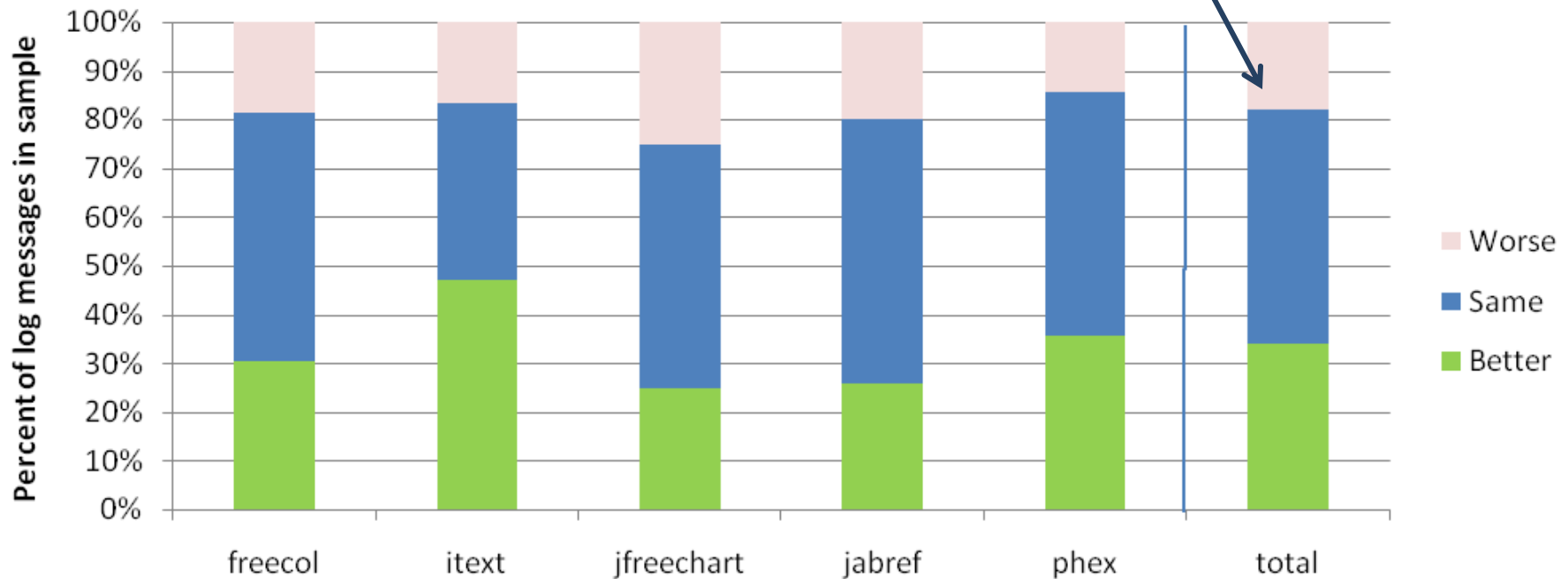


```
When X,  
Do Y  
Instead of Z
```



Evaluation

Our documentation is
as good as human over
80% of the time





API Usage Documentation

“The greatest obstacle to learning an API ... is insufficient or inadequate examples”⁹

9. M. P. Robillard. What Makes APIs Hard to Learn? Answers from Developers. *IEEE Softw.*, 26(6):27-34, 2009.



API Usage Documentation

java.util.ObjectOutputStream

```
FileOutputStream fos = new  
    FileOutputStream("t.tmp");  
ObjectOutputStream oos = new  
    ObjectOutputStream(fos);  
oos.writeInt(12345);  
oos.writeObject("Today");  
oos.writeObject(i);  
oos.close();
```

java.util.BufferedReader

```
BufferedReader in = new  
    BufferedReader(new FileReader("foo.in"));
```

weka.core.Instance

```
// Create the instance  
Instance iExample = new Instance(4);  
iExample.setValue((Attribute)fvWekaAttributes.elementAt(0), 1.0);  
iExample.setValue((Attribute)fvWekaAttributes.elementAt(1), 0.5);  
iExample.setValue((Attribute)fvWekaAttributes.elementAt(2), "gray");  
iExample.setValue((Attribute)fvWekaAttributes.elementAt(3), "positive");  
  
isTrainingSet.add(iExample);
```



Key Idea

- Combine insights from **specification mining**, automatic documentation, and code summarization
- Specification mining *false positives* – usage patterns that are common but aren't required – are exactly what we want to find.



Algorithm

Given a target class to document, and a set of code files that use the class (e.g., mined from the web).

- **Model** usages of the classes as a finite state machine or regular expression
- **Combine** machines that are similar
- Output most common machines as usage examples



Evaluation

Manual comparison to JavaDoc examples

- Are we able to come up with the same examples?
 - Precision / Recall / F-measure
- User Study



Conclusion

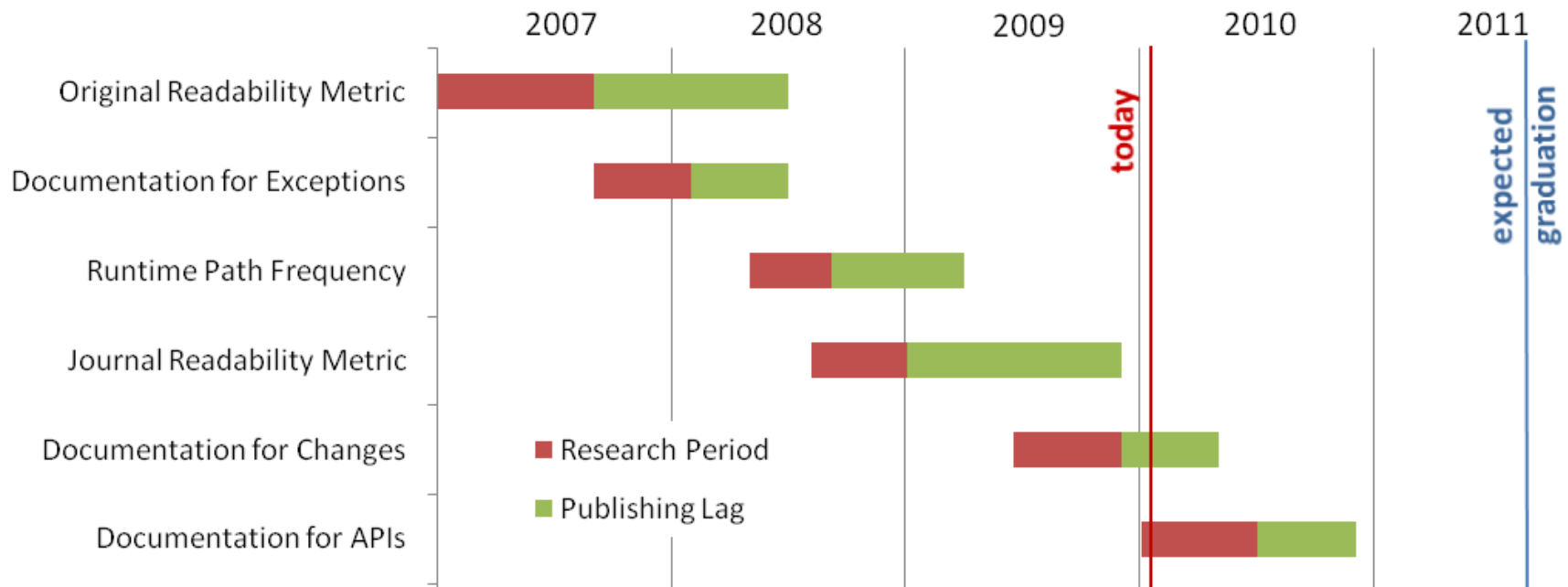
To create algorithms for three types of documentation:

- Exceptions
- Code Changes
- API Usage

Evaluate by comparing to human generated documentation and/or with a user study



Research Timeline





A 2005 NASA survey found that the most significant barrier to **code reuse** is that software is “too difficult to understand” or is “poorly documented.”¹⁰

10. Nasa Software Reuse Working Group. Software reuse survey. http://www.esdswg.com/softwarereuse/Resources/library/working_group_documents/survey2005, 2005.



Conclusion: Understanding programs at many levels

- How easy is it to understand and maintain this software? *Readability*
- Where are the corner cases, and where are the common paths? *Runtime Behavior*
- How can this code go wrong?
Documenting Exceptions
- How do I use this code? *Documenting APIs*
- What does proposed fix really do?
Documenting Changes

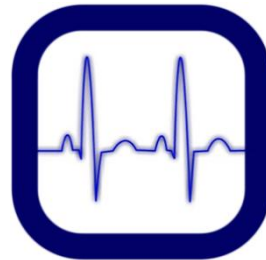
All Questions Encouraged

These slides, the proposal document, and much more information is available at:

<http://arrestedcomputing.com/proposal>



Readability



Runtime Behavior



Documentation

Thanks for Coming!