

# CEN 5016: Software Engineering

Spring 2026



University of  
Central Florida

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Dr. Kevin Moran

## *Week 5 - Class 1:* Static & Dynamic Analysis

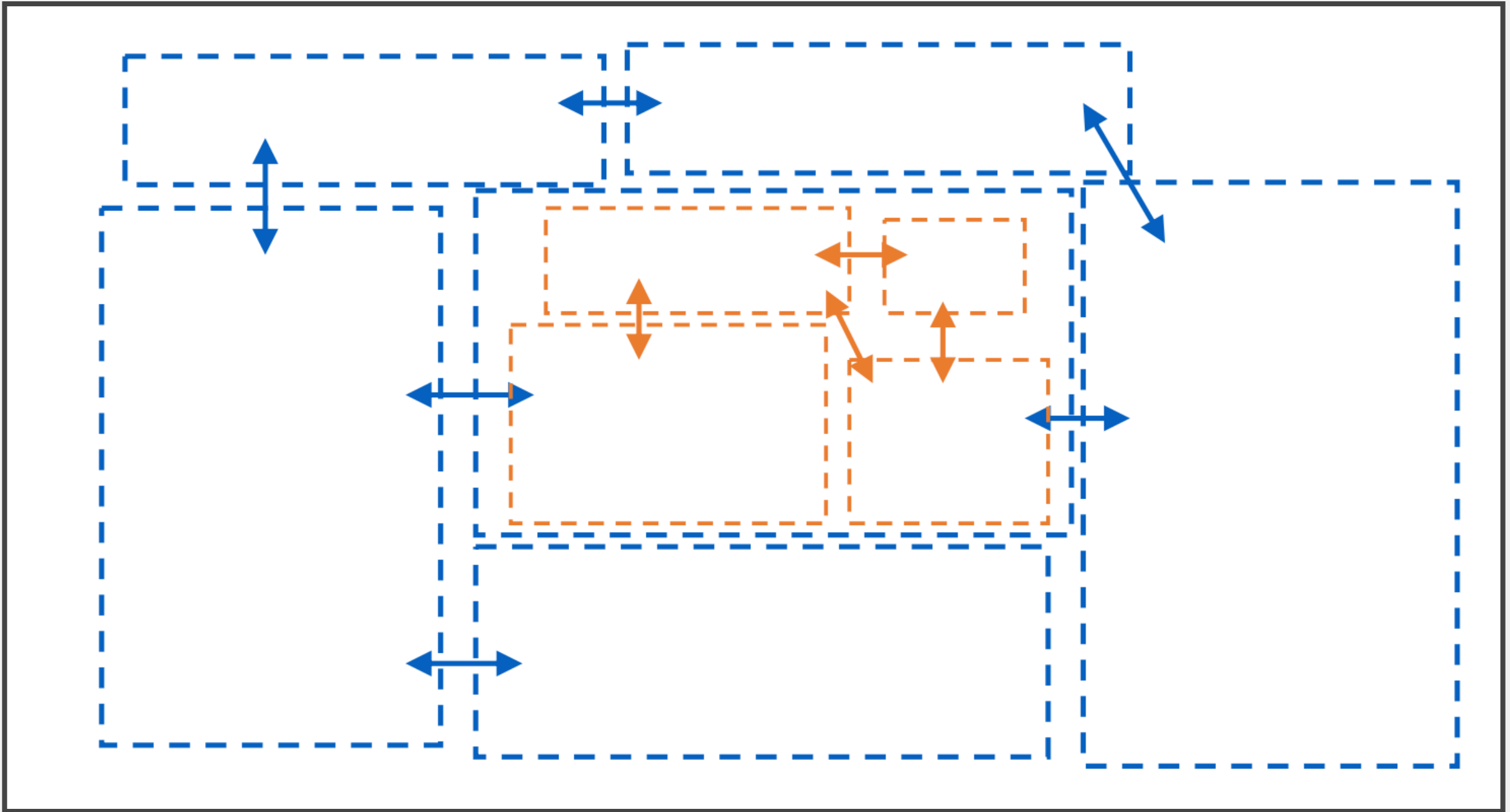




- *Assignment 3*
  - Due tomorrow, Weds, Feb 11th
  - Small Extension due to GitHub issues
- *SDE Project Part 1*
  - Due Friday, Feb 13th
  - Project Teams will be finalized in web courses today!

# Software Architecture







# Why Document Architecture?



- Blueprint for the system
  - Artifact for early analysis
  - Primary carrier of quality attributes
  - Key to post-deployment maintenance and enhancement
- Documentation speaks for the architect, today and 20 years from today
- As long as the system is built, maintained, and evolved according to its documented architecture
- Support traceability.

# Views & Purposes



- Every view should align with a purpose
- Views should only represent information relevant to that purpose
  - Abstract away other details
  - Annotate view to guide understanding where needed
- Different views are suitable for different reasoning aspects (different quality goals), e.g.,
  - Performance
  - Extensibility
  - Security
  - Scalability
  - ...

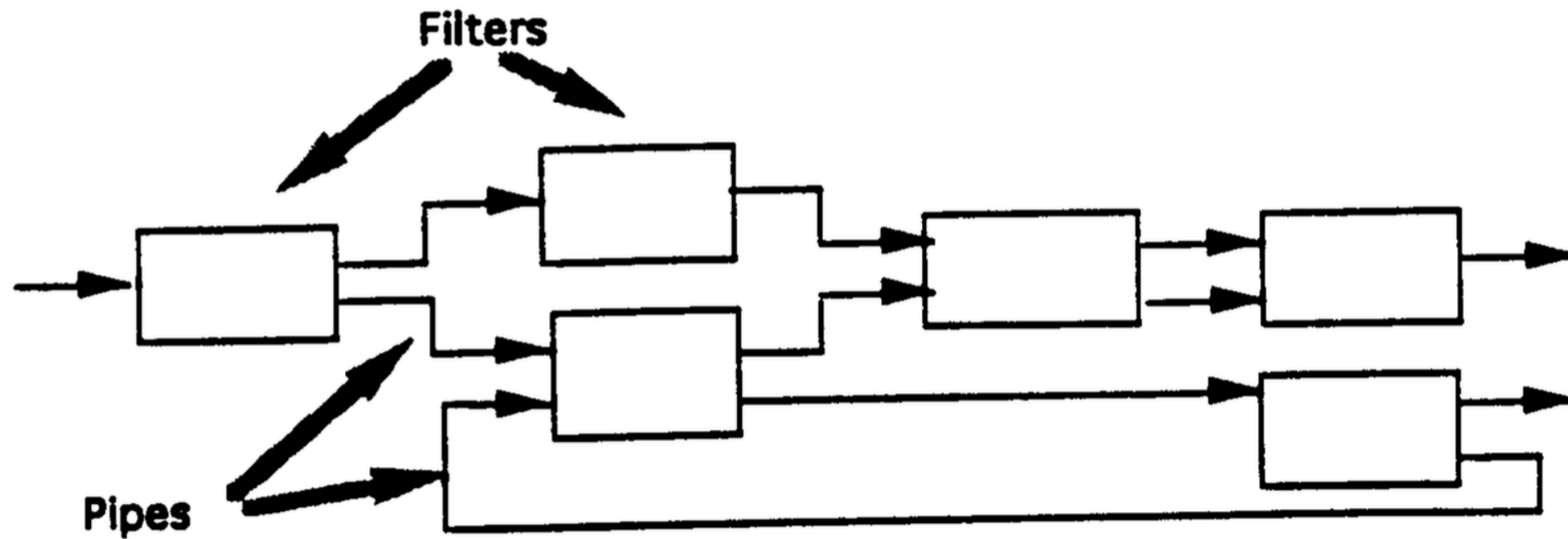


- Static View
  - Modules (subsystems, structures) and their relations (dependencies, ...)
- Dynamic View
  - Components (processes, runnable entities) and connectors (messages, data flow, ...)
- Physical View (Deployment)
  - Hardware structures and their connections

# Common Software Architectures

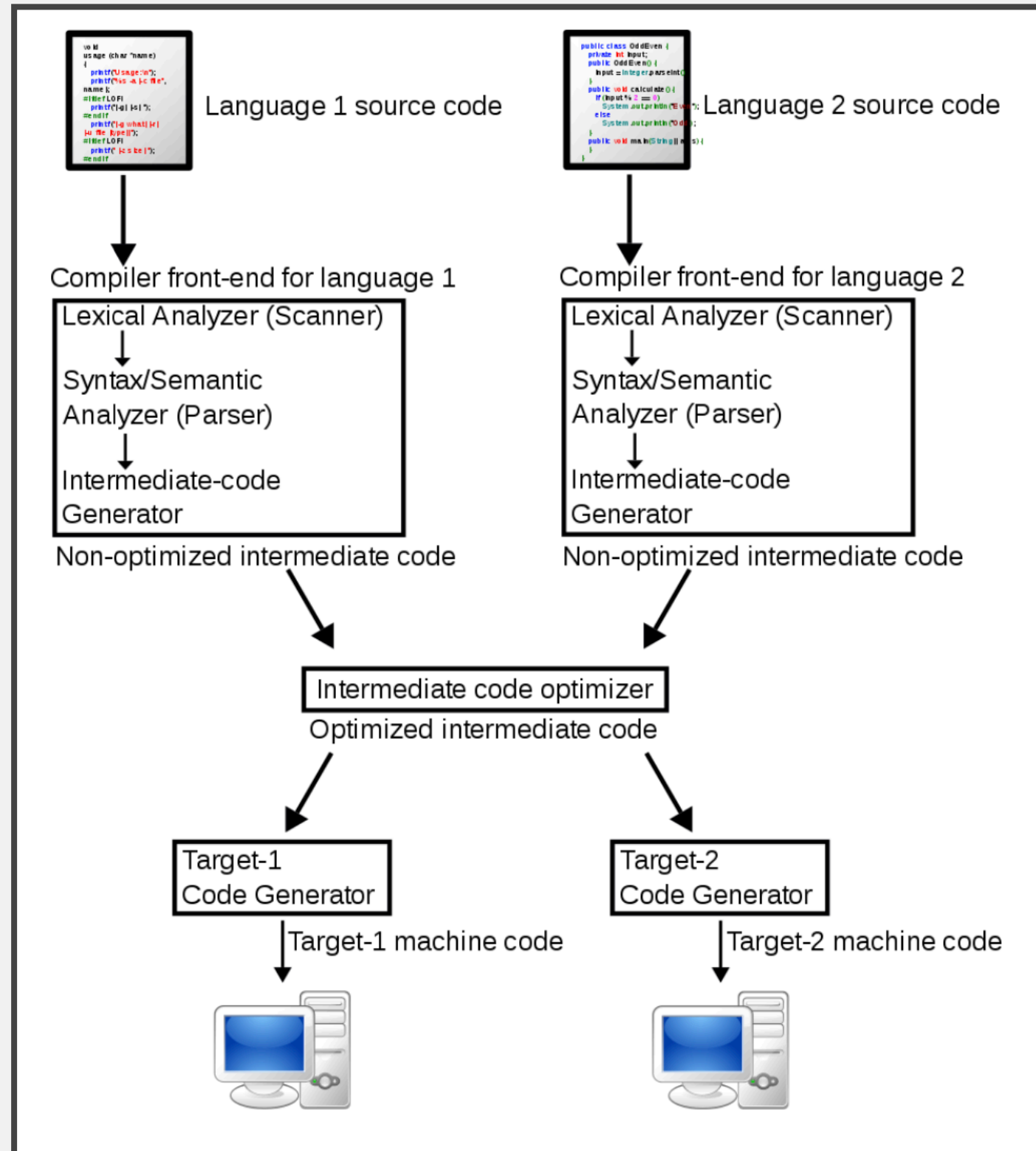


# I. Pipes & Filters

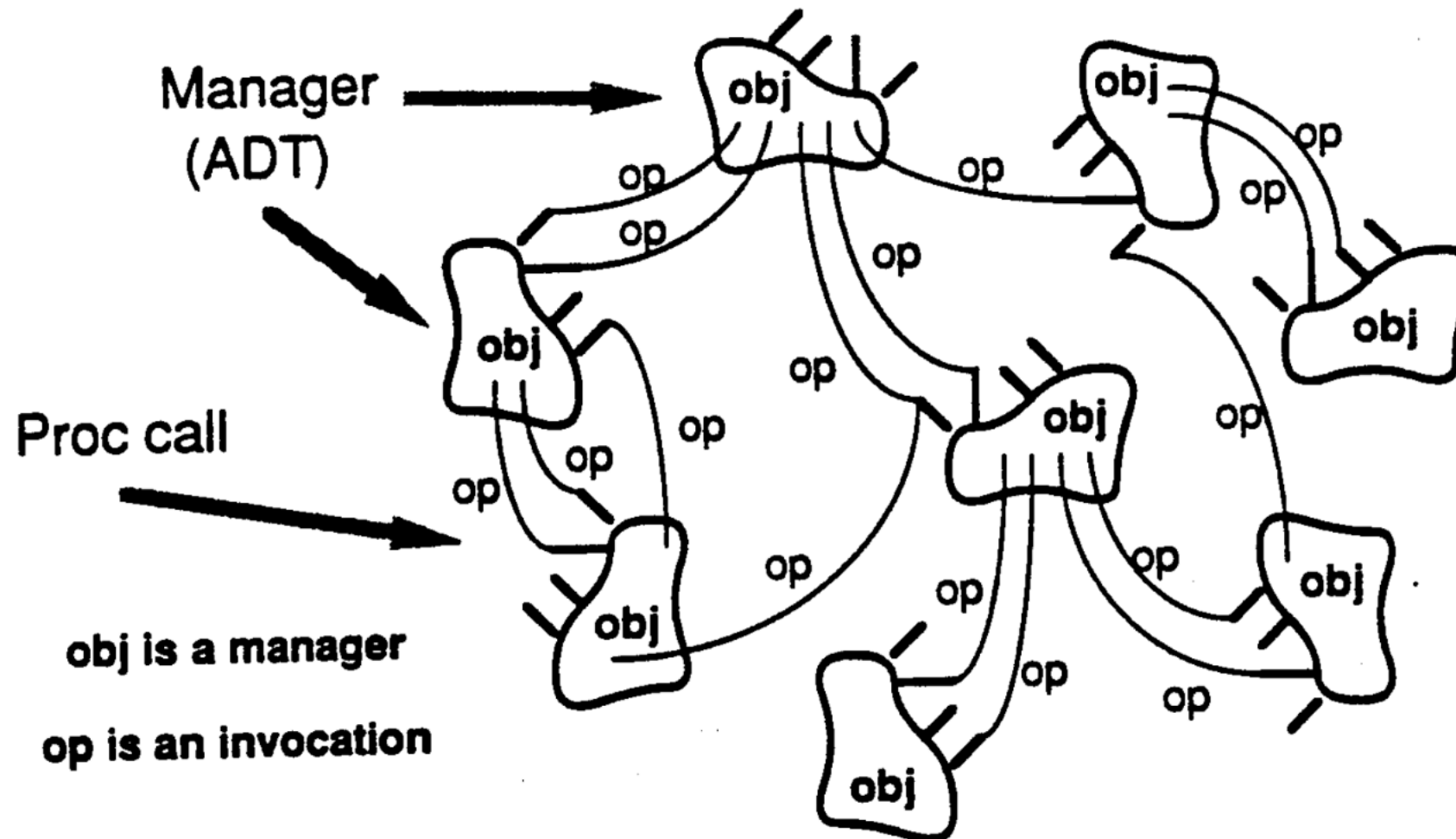


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# Pipes & Filters Example: Compilers

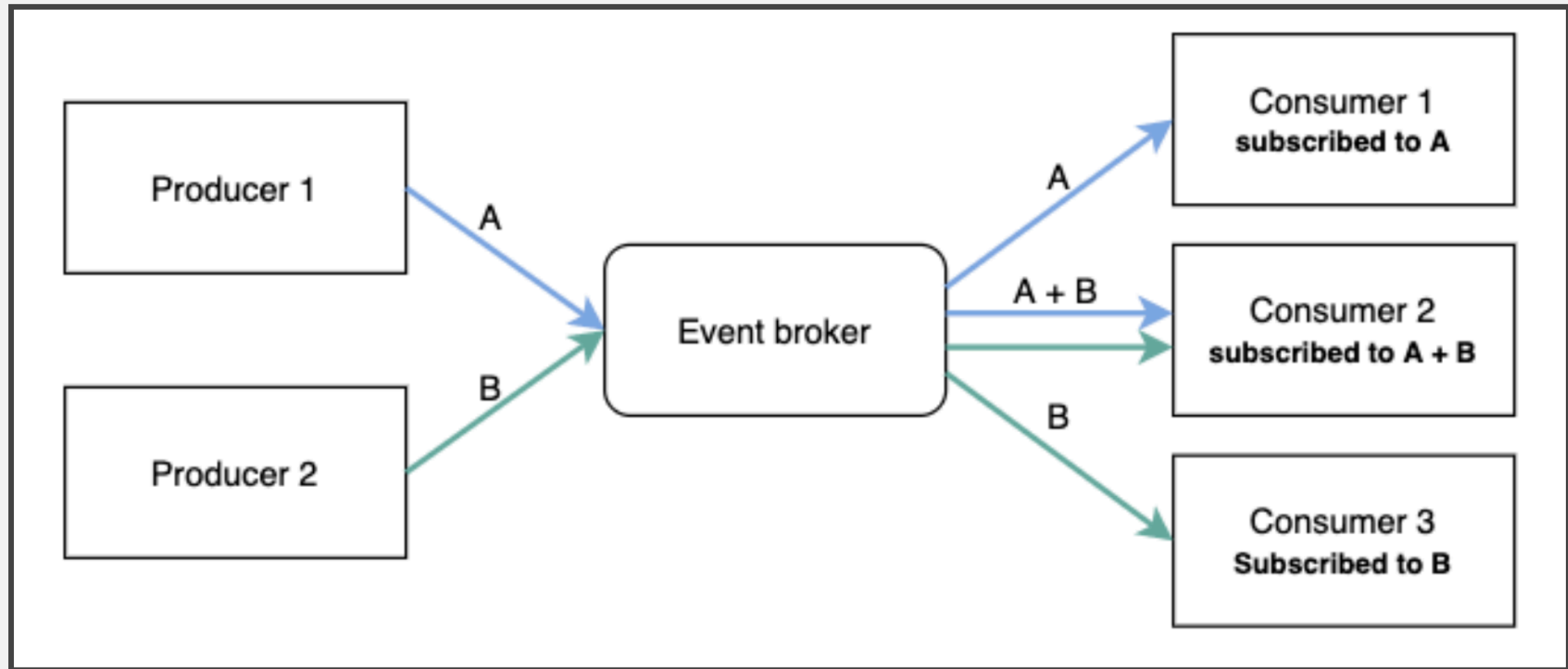


## 2. Object Oriented Organization



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### 3. Event-Driven Architecture

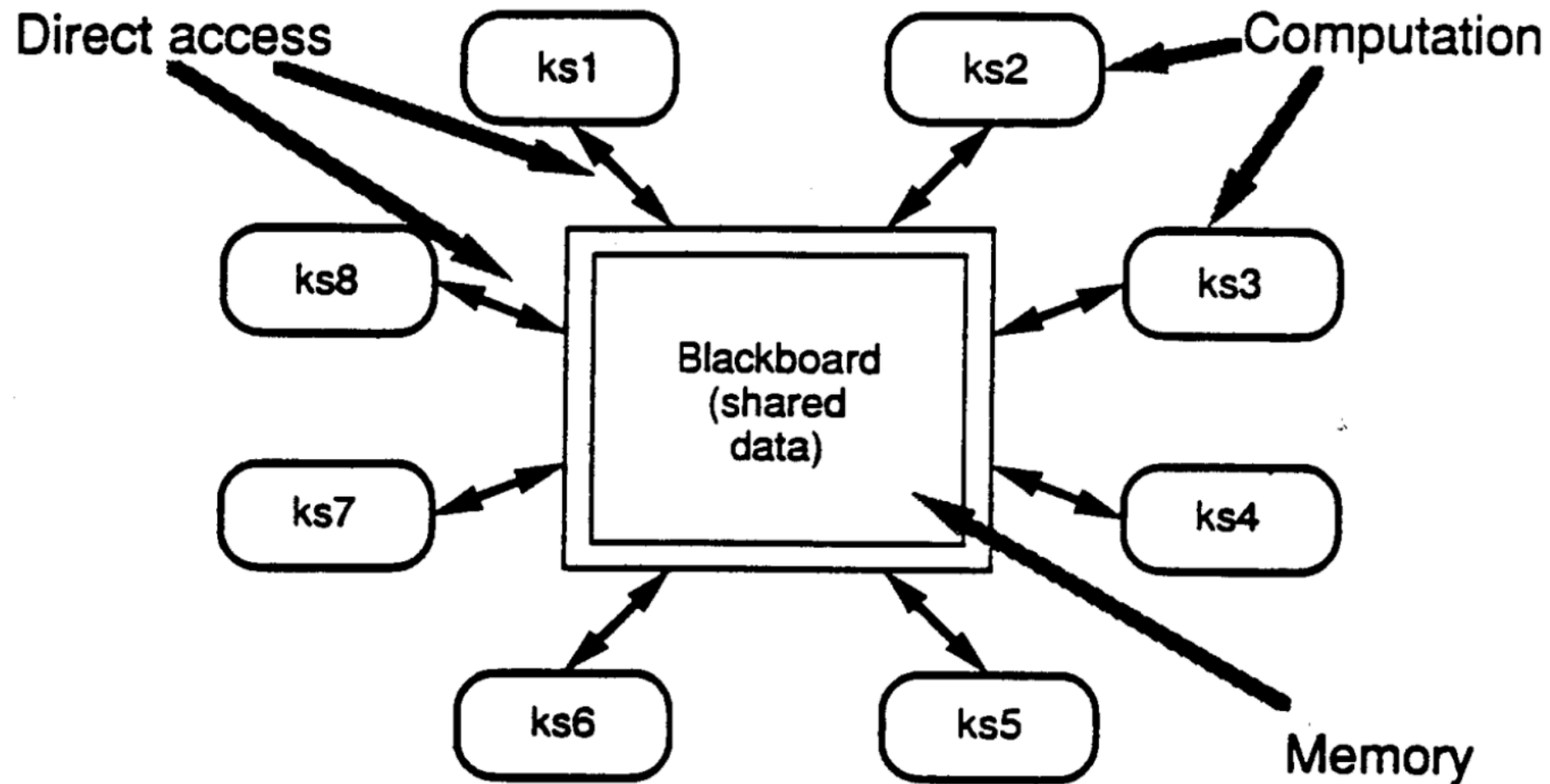






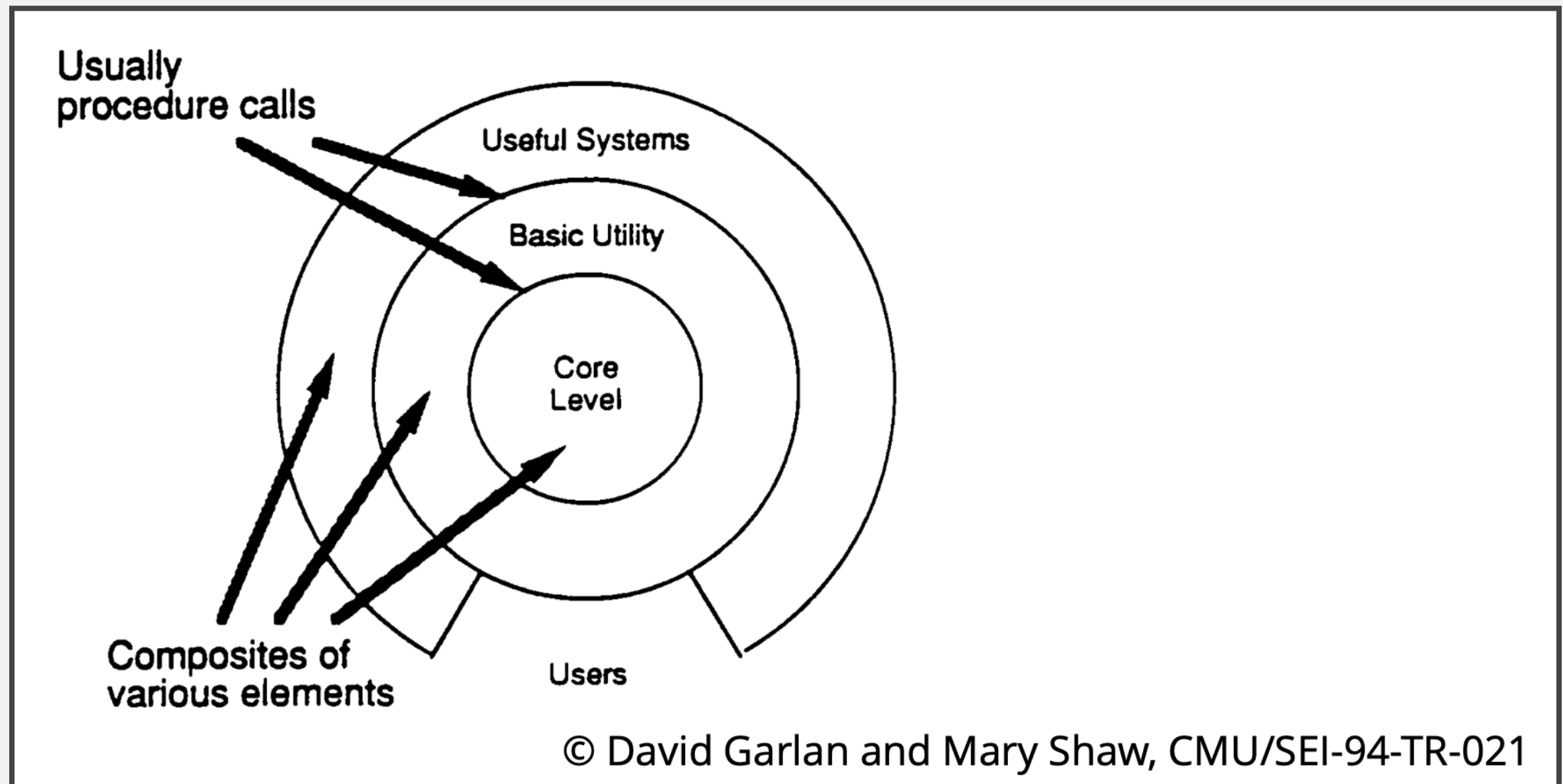
**FAKEFLIX**

## 4. Blackboard Architecture

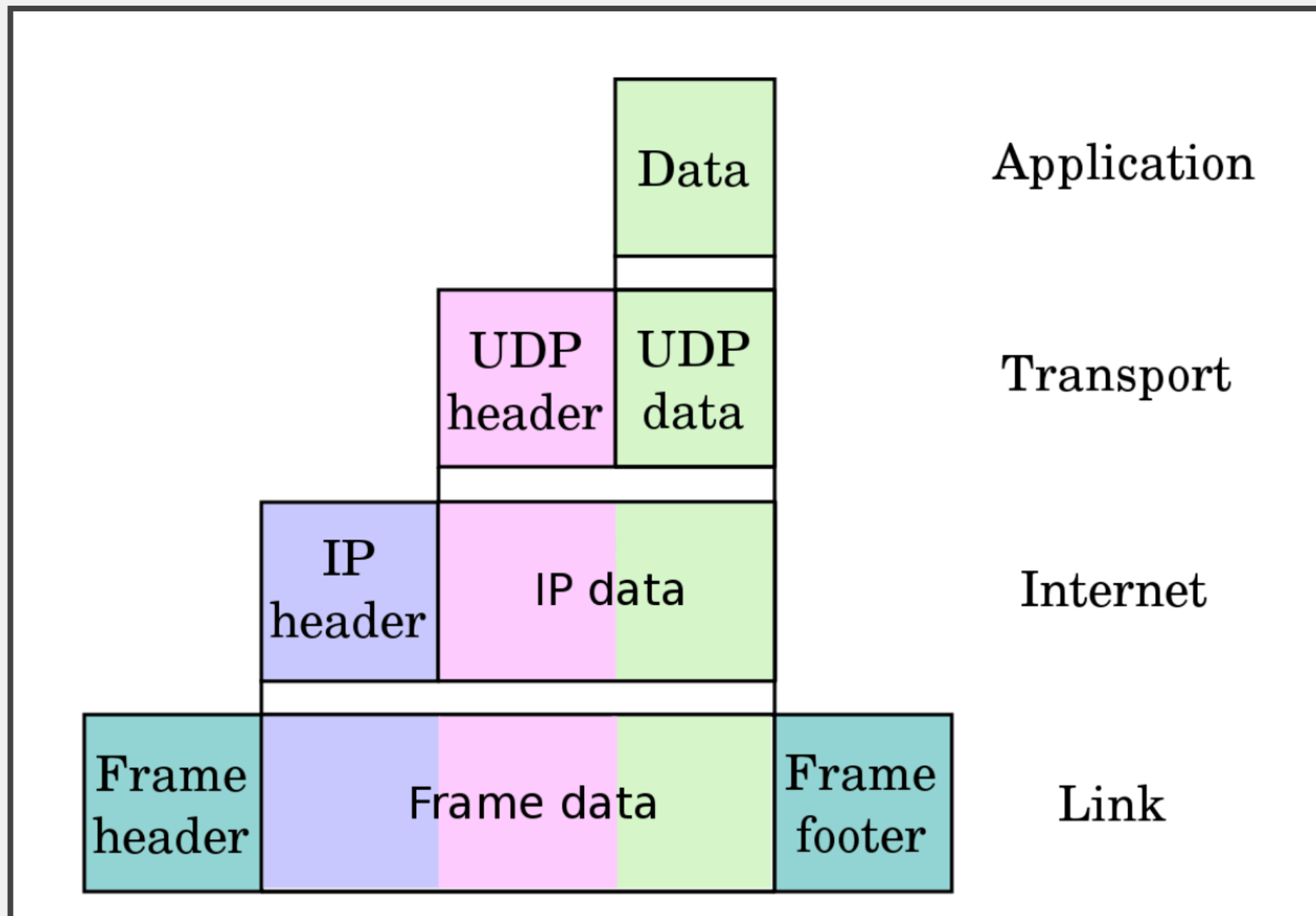


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# 5. Layered Systems



# Example Internet Protocol Suite



# Guidelines for Selecting a Notation



- Suitable for purpose
- Often visual for compact representation
- Usually boxes and arrows
- UML possible (semi-formal), but possibly constraining
  - Note the different abstraction level – Subsystems or processes, not classes or objects
- Formal notations available
- Decompose diagrams hierarchically and in views
- Always include a legend
- Define precisely what the boxes mean
- Define precisely what the lines mean
- Do not try to do too much in one diagram
  - Each view of architecture should fit on a page
  - Use hierarchy

# Software QA: Static & Dynamic Analysis



# Learning Goals



- Gain an understanding of the relative strengths and weaknesses of static and dynamic analysis
- Examine several popular analysis tools and understand their use cases
- Understand how analysis tools are used in large open-source software

# Activity: Analyze the Python Program Statically



```
def n2s(n: int, b: int) -> str:
    if n <= 0:
        return '0'
    r = ""
    while n > 0:
        u = n % b
        if u >= 10:
            u = chr(ord('A') + u - 10)
        n = n // b
        r = str(u) + r
    return r
```

1. What are the set of data types taken by variable `u` at any point in the program?
2. Can the variable `u` be a negative number?
3. Will this function always return a value?
4. Can there ever be a division by zero?
5. Will the returned value ever contain a minus sign '-'?

Answer: Yes, No, Maybe



# What Static Analysis Can & Cannot Do



- Type-checking is well established
  - Set of data types taken by variables at any point
  - Can be used to prevent type errors (e.g. Java) or warn about potential type errors (e.g. Python)
- Checking for problematic patterns in syntax is easy and fast
  - Is there a comparison of two Java strings using ``==``?
  - Is there an array access ``a[i]`` without an enclosing bounds check for ``i``?
- Reasoning about termination is impossible in general
  - Halting problem
- Reasoning about exact values is hard, but conservative analysis via abstraction is possible
  - Is the bounds check before ``a[i]`` guaranteeing that ``i`` is within bounds?
  - Can the divisor ever take on a zero value?
  - Could the result of a function call be ``42``?
  - Will this multi-threaded program give me a deterministic result?
  - Be prepared for “MAYBE”
- Verifying some advanced properties is possible but expensive
  - CI-based static analysis usually over-approximates conservatively

# Bad News: Rice's Theorem



- Every static analysis is necessarily incomplete, unsound, undecidable, or a combination thereof
- *“Any nontrivial property about the language recognized by a Turing machine is undecidable.”*
- Henry Gordon Rice, 1953



- **Security:** Buffer overruns, improperly validated input...
- **Memory safety:** Null dereference, uninitialized data...
- **Resource leaks:** Memory, OS resources...
- **API Protocols:** Device drivers; real time libraries; GUI frameworks
- **Exceptions:** Arithmetic/library/user-defined
- **Encapsulation:**
  - Accessing internal data, calling private functions...
- **Data races:**
  - Two threads access the same data without synchronization



- **Linters**
  - Shallow syntax analysis for enforcing code styles and formatting
- **Pattern-based bug detectors**
  - Simple syntax or API-based rules for identifying common programming mistakes
- **Type-annotation validators**
  - Check conformance to user-defined types
  - Types can be complex (e.g., “Nullable”)
- **Data-flow analysis / Abstract interpretation)**
  - Deep program analysis to find complex error conditions (e.g., “can array index be out of bounds?”)



- Find bugs
- Refactor code
- Keep your code stylish!
- Identify code smells
- Measure quality
- Find usability and accessibility issues
- Identify bottlenecks and improve performance

# Activity: Analyze the Python Program Dynamically



```
def n2s(n: int, b: int) -> str:
    if n <= 0:
        return '0'
    r = ""
    while n > 0:
        u = n % b
        if u >= 10:
            u = chr(ord('A') + u - 10)
        n = n // b
        r = str(u) + r
    return r
print(n2s(12, 10))
```

Answer: Yes, No, Maybe

1. What are the set of data types taken by variable `u` at any point in the program?
2. Did the variable `u` ever contain a negative number?
3. For how many loop executions did the while loop execute?
4. Was there a division by zero?
5. Did the returned value ever contain a minus sign '-'?



- Tells you properties of the program that were definitely observed
  - Code coverage
  - Performance profiling
  - Type profiling
  - Testing
- In practice, implemented by program instrumentation
  - Think “Automated logging”
  - Slows down execution speed by a small amount

# Static Analysis vs. Dynamic Analysis



- Requires only source code
- Conservatively reasons about all possible
- Reported warnings may contain false positives
- Can report all warnings of a particular class of problems
- Advanced techniques like verification can prove certain complex properties, but rarely run in CI due to cost

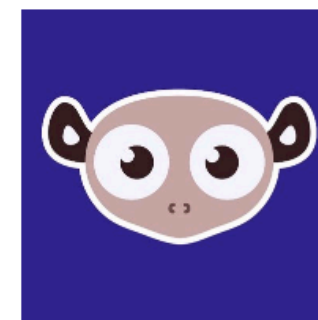
- Requires successful build + test inputs
- Observes individual executions
- Reported problems are real, as observed by a witness input
- Can only report problems that are seen. Highly dependent on test inputs. Subject to false negatives
- Advanced techniques like symbolic execution can prove certain complex properties, but rarely run in CI due to cost



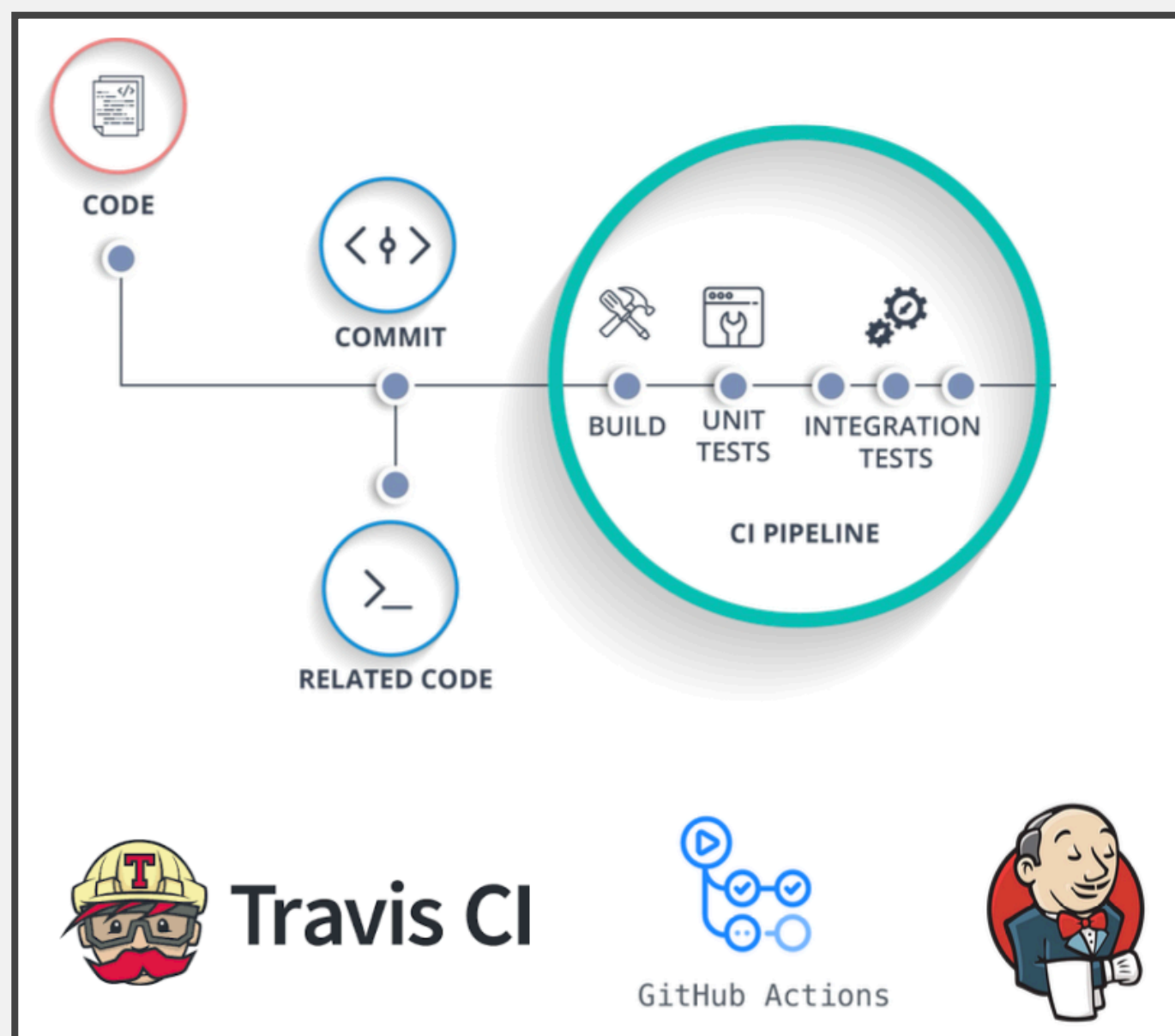
# Static Analysis



# Tools for Static Analysis



# Static Analysis is a Key Part of CI



# Static Analysis used to be Purely Academic...



## GitHub acquires code analysis tool Semmle

Frederic Lardinois @frederici / 1:30 pm EDT • September 18, 2019

Comment



Marketplace Search results

Types

Apps

Actions

Categories

API management  
Chat  
Code quality  
Code review  
Continuous integration  
Dependency management  
Deployment  
IDEs  
Learning  
Localization  
Mobile  
Monitoring  
Project management  
Publishing

Search for apps and actions

Apps

Build on your workflow with apps that integrate with GitHub.

306 results filtered by Apps



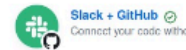
**Zube**  
Agile project management that lets the entire team work with developers on GitHub



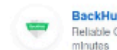
**WhiteSource Bolt**  
Detect open source vulnerabilities in real time with suggested fixes for quick remediation



**Crowdin**  
Agile localization for your projects



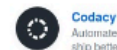
**Slack + GitHub**  
Connect your code without leaving Slack



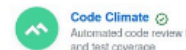
**BackHub**  
Reliable GitHub repository backup, set up in minutes



**GitLocalize**  
Continuous Localization for GitHub projects



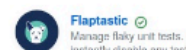
**Codacy**  
Automated code reviews to help developers ship better software, faster



**Code Climate**  
Automated code review for technical debt and test coverage



**Semaphore**  
Test and deploy at the push of a button



**Flaplastic**  
Manage flaky unit tests. Click a checkbox to instantly disable any test on all branches. Works with your current test suite



**DeepScan**  
Advanced static analysis for automatically finding runtime errors in JavaScript code



**Depfu**  
Automated dependency updates done right



News

## Snyk Secures \$150M, Snags \$1B Valuation



**Sydney Sawaya** | Associate Editor  
January 21, 2020 1:12 PM

Share this article:



Snyk, a developer-focused security startup that identifies vulnerabilities in open source applications, announced a \$150 million Series C funding round today. This brings the company's total investment to \$250 million alongside reports that put the company's valuation at more than \$1 billion.



# Static Analysis is Also Integrated into IDEs



```
cppcoreguidelines.cpp x
1 // To enable only C++ Core Guidelines checks
2 // go to Settings/Preferences | Editor | Inspections | C/C++ | Clang-Tidy
3 // and provide: -*,cppcoreguidelines-* in options
4
5 void fill_pointer(int* arr, const int num) {
6     for(int i = 0; i < num; ++i) {
7         arr[i] = 0;
8     }
9     Do not use pointer arithmetic
10
11 void fill_array(int ind) {
12     int arr[3] = {1,2,3};
13     arr[ind] = 0;
14 }
15
16 void cast_away_const(const int& magic_num)
17 {
18     const_cast<int&>(magic_num) = 42;
19 }
20
```

```
97 new Todo({
98   content: item,
99   updated_at: Date.now(),
100 })
101 .save(function (err, todo, count) {
102   if (err) return next(err);
103
104   //
105   res.setHeader('Data', todo.content.toString('base64'));
106   res.redirect('/');
107 }
108
109 res.setHeader('Location', '/');
110 res.status(302).send(todo.content.toString('base64'));
111
112 // res.redirect('/' + todo.content.toString('base64'));
113 });
114
```

**H Cross-site Scripting (XSS)**  
Vulnerability CWE-79

Unsanitized input from the HTTP request body flows into send, where it is used to render an HTML page returned to the user. This may lead to a Cross-site Scripting attack (XSS).

**Data Flow - 12 steps**

```
1 index.js:80 | var item = req.body.content;
2 index.js:80 | if (typeof item !== 'string' && item.match(regex)) {
3 index.js:9 | Click to show in the Editor
4 index.js:55 | function parse(todo) {
5 index.js:56 | var t = todo;
6 index.js:59 | var reminder = t.toString().indexOf(reminderToken);
7 index.js:61 | var time = t.slice(reminder + reminderToken.length);
8 index.js:69 | t = t.slice(0, reminder);
9 index.js:74 | return t;
```

# What Makes a Good Static Analysis Tool?

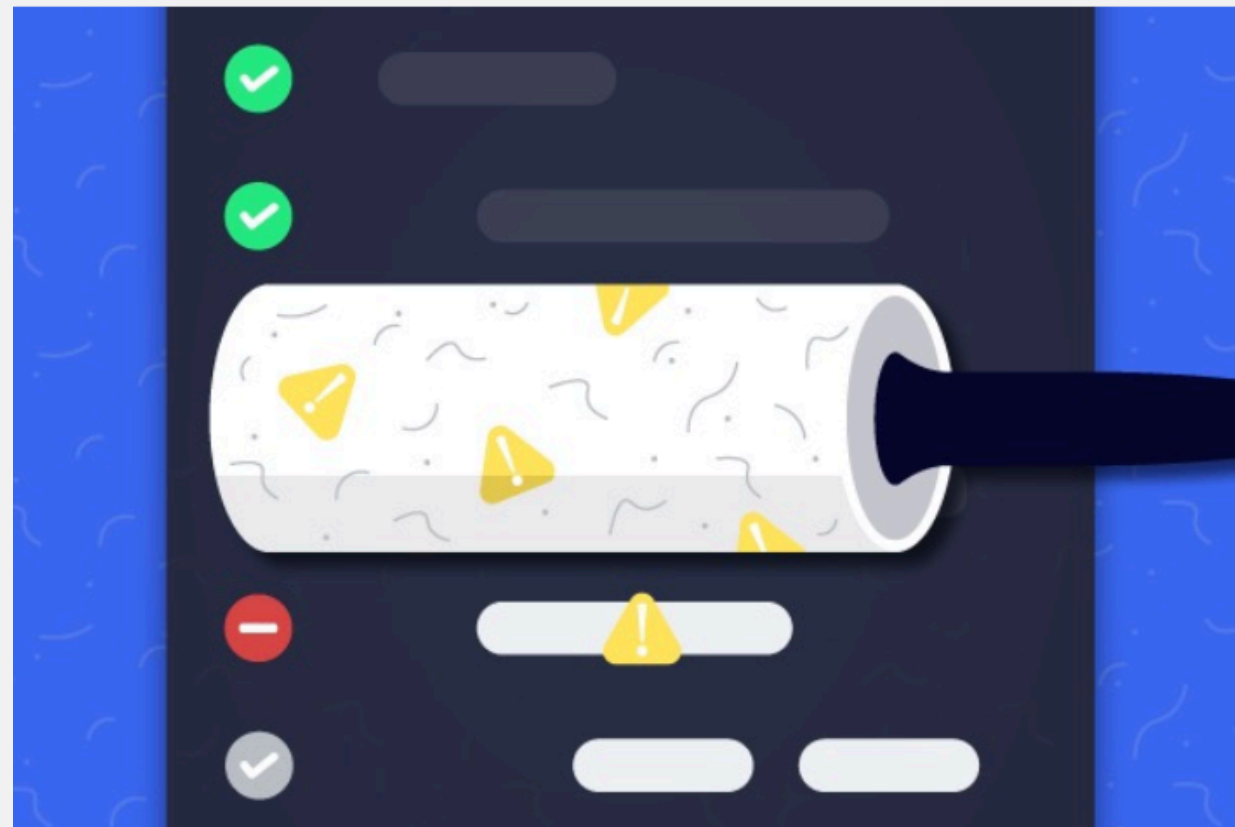


- Static analysis should be fast
  - Don't hold up development velocity
  - This becomes more important as code scales
- Static analysis should report few false positives
  - Otherwise developers will start to ignore warnings and alerts, and quality will decline
- Static analysis should be continuous
  - Should be part of your continuous integration pipeline
  - Diff-based analysis is even better -- don't analyse the entire codebase; just the changes
- Static analysis should be informative
  - Messages that help the developer to quickly locate and address the issue
  - Ideally, it should suggest or automatically apply fixes

# (I) Linters



- Cheap, fast, and lightweight static source analysis





# Use Linters to Enforce Style Guidelines



- Don't rely on manual inspection during code review!

Don't rely on manual inspection during code review!



RuboCop





# Linters Use Very “Shallow” Static Analysis



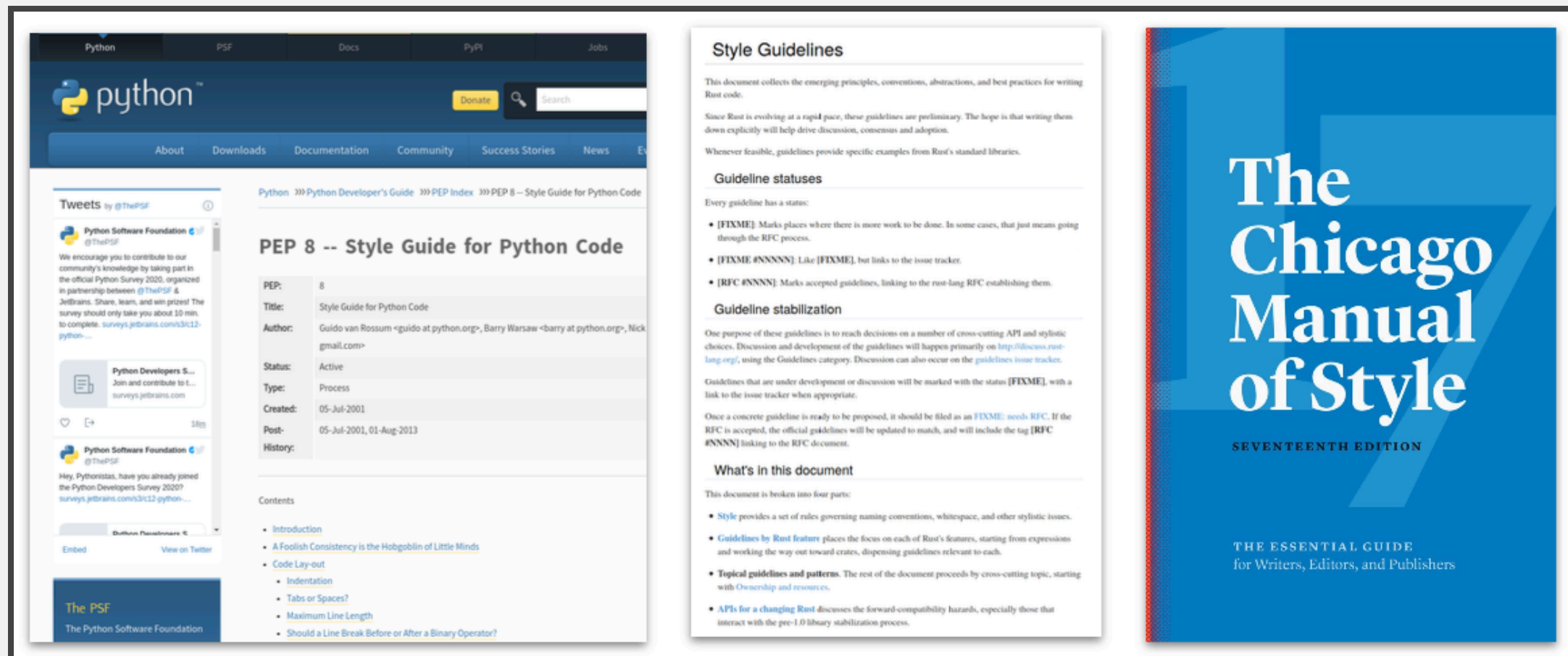
- Ensure proper indentation
- Naming convention
- Line sizes
- Class nesting
- Documenting public functions
- Parenthesis around expressions
- What else?

# Use Linters to Improve Maintainability



- Why? We spend more time reading code than writing it.
  - Various estimates of the exact %, some as high as 80%
- Code ownership is usually shared
- The original owner of some code may move on
- Code conventions make it easier for other developers to quickly understand your code

# UseStyle Guidelines to Facilitate Communication



- Guidelines are inherently opinionated, but consistency is the important point. Agree to a set of conventions and stick to them.

# Take Home Message: Style is an Easy Way to Improve Readability!



- Everyone has their own opinion (e.g., tabs vs. spaces)
- Agree to a convention and stick to it
  - Use continuous integration to enforce it
- Use automated tools to fix issues in existing code

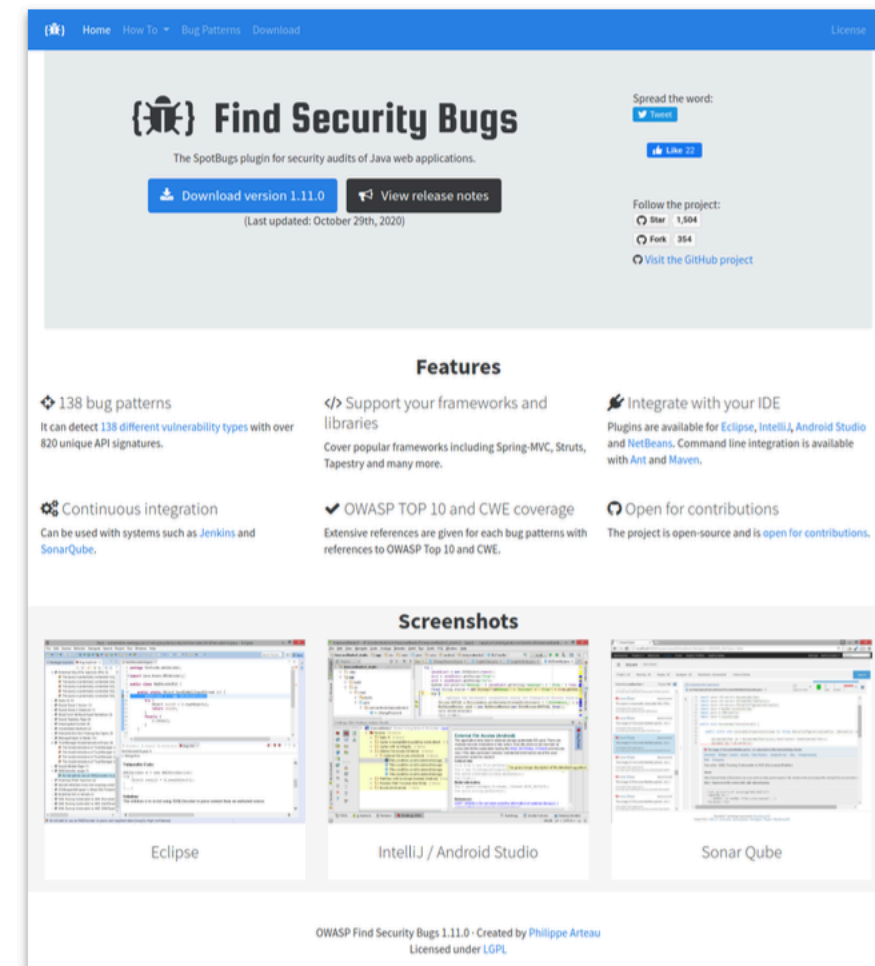
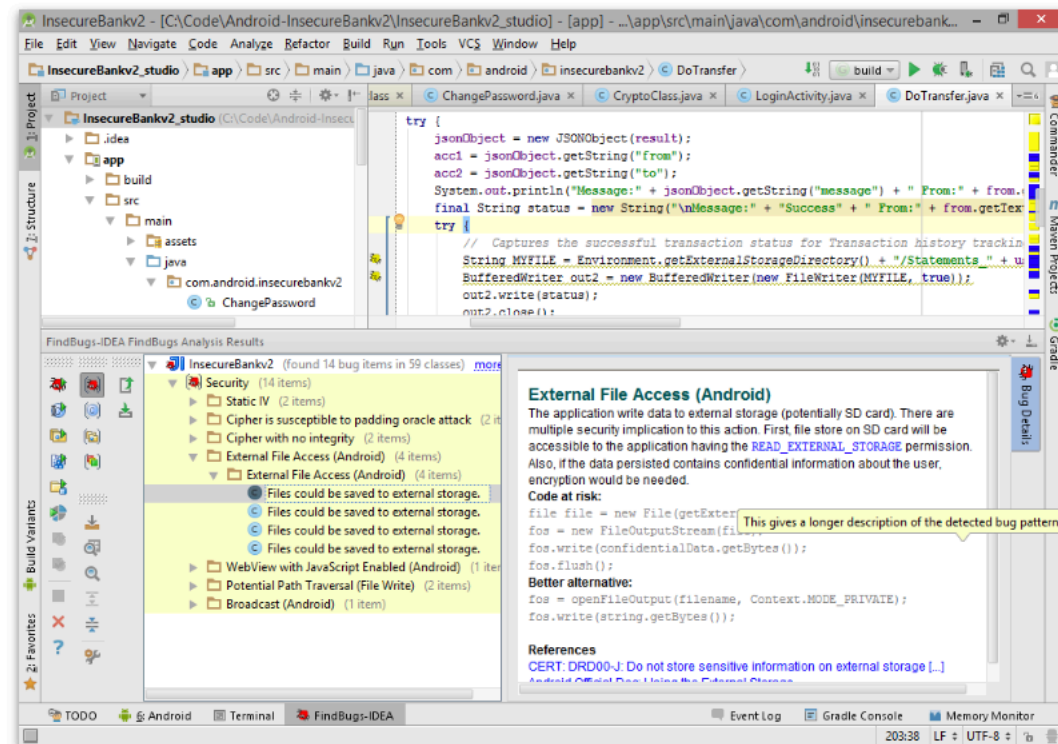
## (2) - Pattern-based Static Analysis Tools



- Bad Practice
- Correctness
- Performance
- Internationalization
- Malicious Code
- Multithreaded Correctness
- Security
- Dodgy Code

FindBugs Bug Descriptions		
This document lists the standard bug patterns reported by FindBugs version 3.0.1.		
<b>Summary</b>		
Description	Category	
BC: Equals method should not assume anything about the type of its argument	Bad practice	
BIT: Check for sign of bitwise operation	Bad practice	
CN: Class implements Cloneable but does not define or use clone method	Bad practice	
CN: clone method does not call super.clone()	Bad practice	
CN: Class defines clone() but doesn't implement Cloneable	Bad practice	
CNT: Rough value of known constant found	Bad practice	
Co: Abstract class defines covariant compareTo() method	Bad practice	
Co: compareTo()/compare() incorrectly handles float or double value	Bad practice	
Co: compareTo()/compare() returns Integer.MIN_VALUE	Bad practice	
Co: Covariant compareTo() method defined	Bad practice	
DE: Method might drop exception	Bad practice	
DE: Method might ignore exception	Bad practice	
DMI: Adding elements of an entry set may fail due to reuse of Entry objects	Bad practice	
DMI: Random object created and used only once	Bad practice	
DMI: Don't use removeAll to clear a collection	Bad practice	
Dm: Method invokes System.exit(...)	Bad practice	
Dm: Method invokes dangerous method runFinalizersOnExit	Bad practice	
ES: Comparison of String parameter using == or !=	Bad practice	
ES: Comparison of String objects using == or !=	Bad practice	
Eq: Abstract class defines covariant equals() method	Bad practice	
Eq: Equals checks for incompatible operand	Bad practice	
Eq: Class defines compareTo(...) and uses Object.equals()	Bad practice	
Eq: equals method fails for subtypes	Bad practice	
Eq: Covariant equals() method defined	Bad practice	
FI: Empty finalizer should be deleted	Bad practice	
FI: Explicit invocation of finalizer	Bad practice	
FI: Finalizer nulls fields	Bad practice	
FI: Finalizer only nulls fields	Bad practice	
FI: Finalizer does not call superclass finalizer	Bad practice	
FI: Finalizer nullifies superclass finalizer	Bad practice	
FI: Finalizer does nothing but call superclass finalizer	Bad practice	
FS: Format string should use %n rather than \n	Bad practice	
GC: Unchecked type in generic call	Bad practice	
HE: Class defines equals() but not hashCode()	Bad practice	
HE: Class defines equals() and uses Object.hashCode()	Bad practice	
HE: Class defines hashCode() but not equals()	Bad practice	
HE: Class defines hashCode() and uses Object.equals()	Bad practice	
HE: Class inherits equals() and uses Object.hashCode()	Bad practice	
IC: Superclass uses subclass during initialization	Bad practice	
IMSE: Dubious catching of IllegalStateException	Bad practice	
ISC: Needless instantiation of class that only supplies static methods	Bad practice	
It: Iterator.next() method can't throw NoSuchElementException	Bad practice	
JZEE: Store of non-serializable object into HttpSession	Bad practice	
JCIP: Fields of immutable classes should be final	Bad practice	
ME: Public enum method unconditionally sets its field	Bad practice	

# SpotBugs can be Extended with Plugins





- The analysis must produce zero false positives
  - Otherwise developers won't be able to build the code!
- The analysis needs to be really fast
  - Ideally  $< 100$  ms
  - If it takes longer, developers will become irritated and lose productivity
- You can't just "turn on" a particular check
  - Every instance where that check fails will prevent existing code from
  - There could be thousands of violations for a single check across large codebases



### (3) -Use Type Annotations to Detect Common Errors



- Uses a conservative analysis to prove the absence of certain defects
- Null pointer errors, uninitialized fields, certain liveness issues, information leaks, SQL injections, bad regular expressions, incorrect physical units, bad format strings, ...
- C.f. SpotBugs which makes no safety guarantees
- Assuming that code is annotated and those annotations are correct
- Uses annotations to enhance type system
- Example: Java Checker Framework or MyPy





### (3) -Use Type Annotations to Detect Common Errors



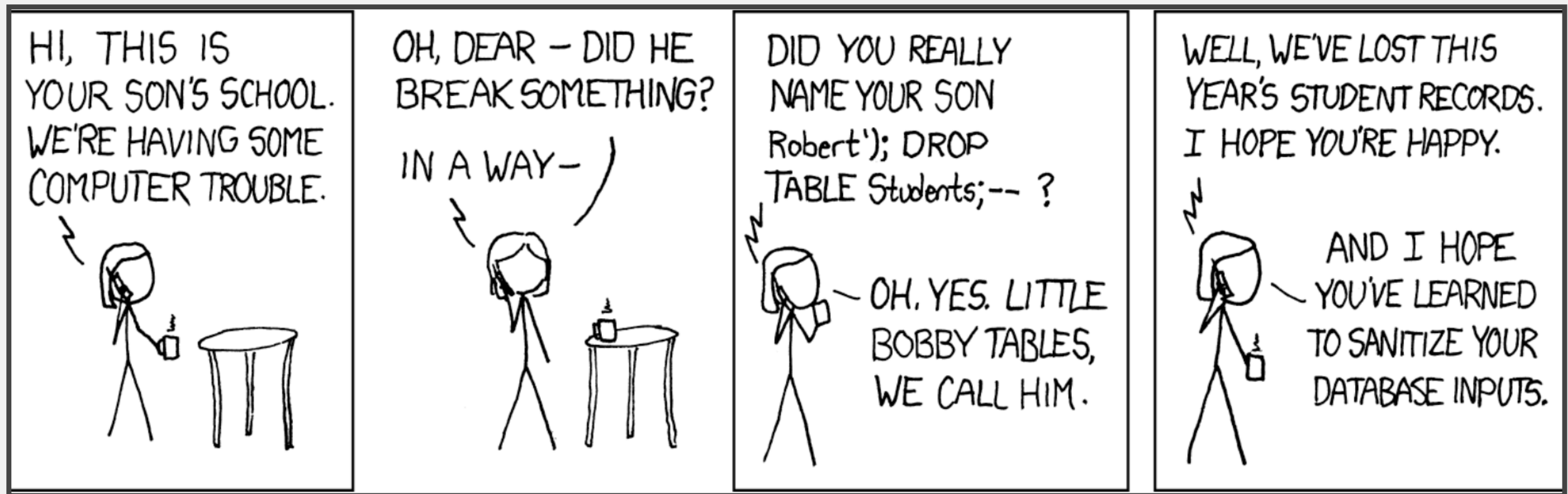
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- Example: Java Checker Framework or MyPy





- Tracks flow of sensitive information through the program
- Tainted inputs come from arbitrary, possibly malicious sources
  - User inputs, unvalidated data
- Using tainted inputs may have dangerous consequences
  - Program crash, data corruption, leak private data, etc.
- We need to check that inputs are sanitized before reaching sensitive locations

# Classic Example: SQL Injection



# Classic Example: SQL Injection



```
void processRequest() {  
    String input = getUserInput();  
    String query = "SELECT ... " + input;  
    executeQuery(query);  
}
```

# Classic Example: SQL Injection



```
void processRequest() {  
    String input = getUserInput();  
    String query = "SELECT ... " + input;  
    executeQuery(query);  
}
```

Tainted input arrives from untrusted source

Tainted output flows to a sensitive sink

# Classic Example: SQL Injection



```
void processRequest() {  
    String input = getUserInput();  
    input = sanitizeInput(input);  
    String query = "SELECT ..." + input;  
    executeQuery(query);  
}
```

Taint is removed by sanitizing data

We can now safely execute query on untainted data



**Remember the Mars Climate Orbiter incident from 1999?**

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## When NASA Lost a Spacecraft Due to a Metric Math Mistake

WRITTEN BY: Ajay Harish | UPDATED ON: March 10th, 2020 | APPROX. READING TIME: 11 Minutes

Blog > CAE Hub > When NASA Lost a Spacecraft Due to a Metric Math Mistake

In September of 1999, after almost 10 months of travel to Mars, the Mars Climate Orbiter burned and broke into pieces. On a day when NASA engineers were expecting to celebrate, the ground reality turned out to be completely different, all because someone failed to use the right units, i.e., the metric units! The Scientific American Space Lab made a brief but interesting video on this very topic.

NASA'S LOST SPACECRAFT

### The Metric System and NASA's Mars Climate Orbiter

The Mars Climate Orbiter, built at a cost of \$125 million, was a 338-kilogram robotic space probe launched by NASA on December 11, 1998 to study the Martian climate, Martian atmosphere, and surface changes. In addition, its function was to act as the communications relay in the Mars Surveyor '98 program for the Mars Polar Lander. The navigation team at the Jet Propulsion Laboratory (JPL) used the metric system of millimeters and meters in its calculations, while

**NASA's Mars Climate Orbiter (cost of \$327 million) was lost because of a discrepancy between use of metric unit Newtons and imperial measure Pound-force.**



- Guarantees that operations are performed on the same kinds and units
- Kinds of annotations
  - @Acceleration, @Angle, @Area, @Current, @Length, @Luminance, @Mass, @Speed, @Substance, @Temperature, @Time
- SI unit annotation
  - @m, @km, @mm, @kg, @mPERs, @mPERs2, @radians, @degrees, @A, ...





- Can only analyze code that is annotated
  - Requires that dependent libraries are also annotated
  - Can be tricky, but not impossible, to retrofit annotations into existing codebases
- Only considers the signature and annotations of methods
  - Doesn't look at the implementation of methods that are being called
- Dynamically generated code
  - Spring Framework
- ● Can produce false positives!
  - Byproduct of necessary approximations

# Infer: What if we didn't want Annotations



- Focused on memory safety bugs
  - Null pointer dereferences, memory leaks, resource leaks, ...
- Compositional interprocedural reasoning
  - Based on separation logic and bi-abduction
- Scalable and fast
  - Can run incremental analysis on changed code
- Does not require annotations
- Supports multiple languages
  - Java, C, C++, Objective-C
  - Programs are compiled to an intermediate representation





## NULLPTR\_DEREFERENCE

Reported as "Nullptr Dereference" by [pulse](#).

Infer reports null dereference bugs in Java, C, C++, and Objective-C when it is possible that the null pointer is dereferenced, leading to a crash.

### Null dereference in Java

Many of Infer's reports of potential Null Pointer Exceptions (NPE) come from code of the form

```
p = foo(); // foo() might return null
stuff();
p.goo();   // dereferencing p, potential NPE
```





## Examples

Infer's cost analysis statically estimates the execution cost of a program without running the code. For instance, assume that we had the following program:

```
void loop(ArrayList<Integer> list){  
  for (int i = 0; i <= list.size(); i++){  
  }  
}
```

For this program, Infer statically infers a polynomial (e.g.  $8|list|+16$ ) for the execution cost of this program by giving each instruction in Infer's intermediate language a symbolic cost (where  $|.|$  refers to the length of a list). Here---overlooking the actual constants---the analysis infers that this program's asymptotic complexity is  $O(|list|)$ , that is loop is linear in the size of its input list. Then, at diff time, if a developer modifies this code to,

# Beware of Inevitable False Positives





openssl / openssl

Sponsor Watch 906 Star 14.2k Fork 6.3k

Code Issues 1.2k Pull requests 251 Actions Projects 2 Wiki Security

Consider using Facebook's "infer" static analysis tool #6968 New issue

Open richsalz opened this issue on Aug 28, 2018

 **dot-asm** commented on Sep 2, 2018 Contributor  ...

I'm not impressed. Majority, >2/3 of reports are DEAD\_STORE and most common reason is last `*ptr++`. More specifically `++` is viewed problematic because *pointer* is not used anymore. The post-increment is also customarily part of macro, so that in order to address this, one would have to have two macros, one that leaves pointer post-incremented and one that doesn't. It would be excessive and doesn't help readability.

Majority of MEMORY\_LEAK reports is because it fails to recognize for example `EVP_MD_CTX_free` as resource freeing. This is counter-productive, one has to work too hard look for real ones. There seem to be couple in `test/*...` Then there is some hairy stuff in `o_names.c:236`, maybe false positive... Oh! There seem to be real leak in `ssl3_final_finish_mac()`, multiple logical errors...



# The Best QA Strategies use Multiple Tools



## How Many of All Bugs Do We Find? A Study of Static Bug Detectors

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### ABSTRACT

Static bug detectors are becoming increasingly popular and are widely used by professional software developers. While most work on bug detectors focuses on whether they find bugs at all, and on how many false positives they report in addition to legitimate warnings, the inverse question is often neglected: How many of all real-world bugs do static bug detectors find? This paper addresses this question by studying the results of applying three widely used static bug detectors to an extended version of the Defects4J dataset that consists of 15 Java projects with 594 known bugs. To decide which of these bugs the tools detect, we use a novel methodology that combines an automatic analysis of warnings and bugs with a manual validation of each candidate of a detected bug. The results of the study show that: (i) static bug detectors find a non-negligible amount of all bugs, (ii) different tools are mostly complementary to each other, and (iii) current bug detectors miss the large majority of the studied bugs. A detailed analysis of bugs missed by the static detectors shows that some bugs could have been found by variants of the existing detectors, while others are domain-specific problems that do not match any existing bug pattern. These findings help potential users of such tools to assess their utility, motivate and outline directions for future work on static bug detection, and provide a basis for future comparisons of static bug detection with other bug finding techniques, such as manual and automated testing.

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### 1 INTRODUCTION

Finding software bugs is an important but difficult task. For average industry code, the number of bugs per 1,000 lines of code has been estimated to range between 0.5 and 25 [21]. Even after years of deployment, software still contains unnoticed bugs. For example, studies of the Linux kernel show that the average bug remains in the kernel for a surprisingly long period of 1.5 to 1.8 years [8, 24]. Unfortunately, a single bug can cause serious harm, even if it has been subsisting for a long time without doing so, as evidenced by examples of software bugs that have caused huge economic losses and even killed people [17, 28, 46].

Given the importance of finding software bugs, developers rely on several approaches to reveal programming mistakes. One approach is to identify bugs during the development process, e.g., through pair programming or code review. Another direction is testing, ranging from purely manual testing over semi-automated testing, e.g., via manually written but automatically executed unit tests, to fully automated testing, e.g., with UI-level testing tools. Once the software is deployed, runtime monitoring can reveal so far missed bugs. e.g., collect information about abnormal runtime

Tool	Bugs
Error Prone	8
Infer	5
SpotBugs	18
Total:	31
Total of 27 unique bugs	

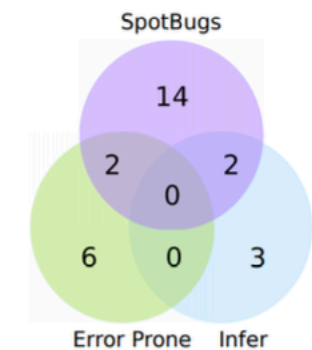


Figure 4: Total number of bugs found by all three static checkers and their overlap.

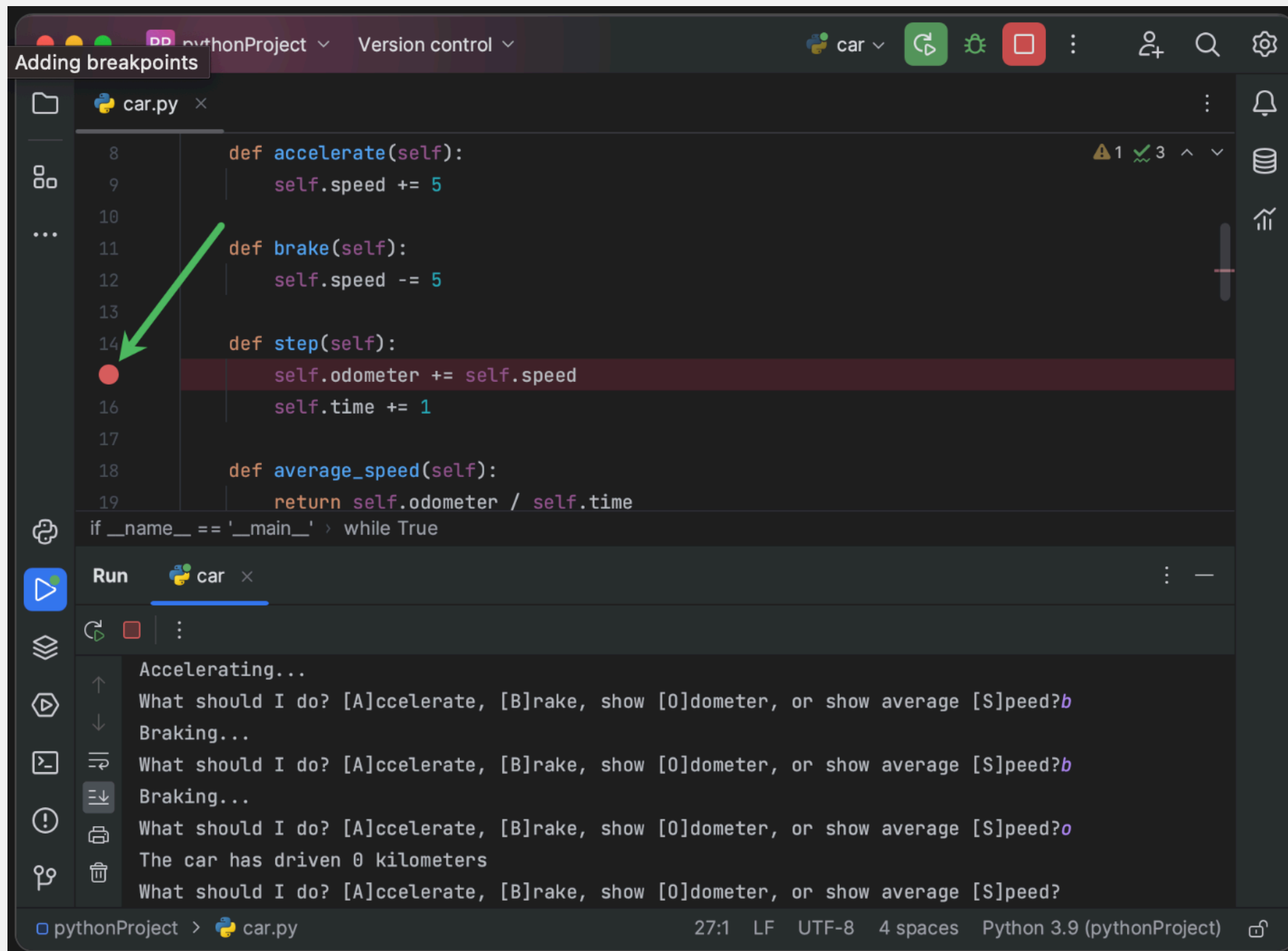
# Dynamic Analysis







# Pycharm Debugger





Valgrind is an instrumentation framework for building dynamic analysis tools. There are Valgrind tools that can automatically detect many memory management and threading bugs, and profile your programs in detail. You can also use Valgrind to build new tools.

The Valgrind distribution currently includes seven production-quality tools: a memory error detector, two thread error detectors, a cache and branch-prediction profiler, a call-graph generating cache and branch-prediction profiler, and two different heap profilers. It also includes an experimental SimPoint basic block vector generator. It runs on the following platforms: X86/Linux, AMD64/Linux, ARM/Linux, ARM64/Linux, PPC32/Linux, PPC64/Linux, PPC64LE/Linux, S390X/Linux, MIPS32/Linux, MIPS64/Linux, X86/Solaris, AMD64/Solaris, ARM/Android (2.3.x and later), ARM64/Android, X86/Android (4.0 and later), MIPS32/Android, X86/FreeBSD, AMD64/FreeBSD, ARM64/FreeBSD, X86/Darwin and AMD64/Darwin (Mac OS X 10.12).

Valgrind is [Open Source](#) / [Free Software](#), and is freely available under the [GNU General Public License, version 2](#).



- Linters are cheap, fast, but imprecise analysis tools
  - Can be used for purposes other than bug detection (e.g., style)
- Conservative analyzers can demonstrate the absence of particular defects
  - At the cost of false positives due to necessary approximations
  - Inevitable trade-off between false positives and false negatives
- The best QA strategy involves multiple analysis and testing techniques
  - The exact set of tools and techniques depends on context