

Combined Static and Dynamic Automated Test Generation

Sai Zhang

University of Washington

Joint work with:

David Saff, Yingyi Bu, Michael D. Ernst

Unit Testing for Object-oriented Programs

- Unit test = sequence of method calls + testing oracle
- Automated test generation is **challenging**:
 - **Legal** sequences for constrained interfaces
 - **Behaviorally-diverse** sequences for good coverage
 - **Testing oracles (assertions)** to detect errors

Unit Testing a Database Program

```
public void testConnection() {  
    Driver driver = new Driver();
```

Constraint 1:
Method-call orders

```
    Connection connection =
```

① → driver.connect("jdbc:tinysql");

② → Statement s = connection.createStatement();

③ → s.execute("create table test (name char(25))");

```
    ....
```

```
    s.close();
```

```
    connection.close();
```

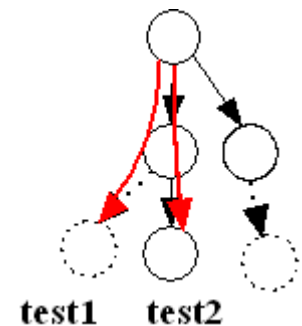
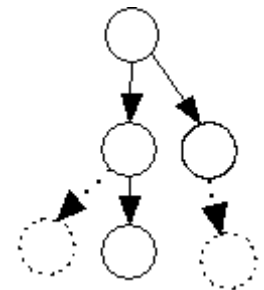
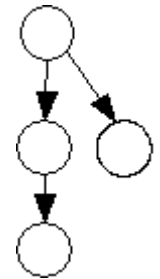
```
}
```

Constraint 2:
Argument values

It is hard to create tests automatically!

Palus: Combining Dynamic and Static Analyses

- **Dynamically** infer an object behavior model from a sample (correct) execution trace
 - Capture method-call order and argument constraints
- **Statically** identify related methods
 - Expand the (incomplete) dynamic model
- **Model-Guided** random test generation
 - Fuzz along a specific legal path

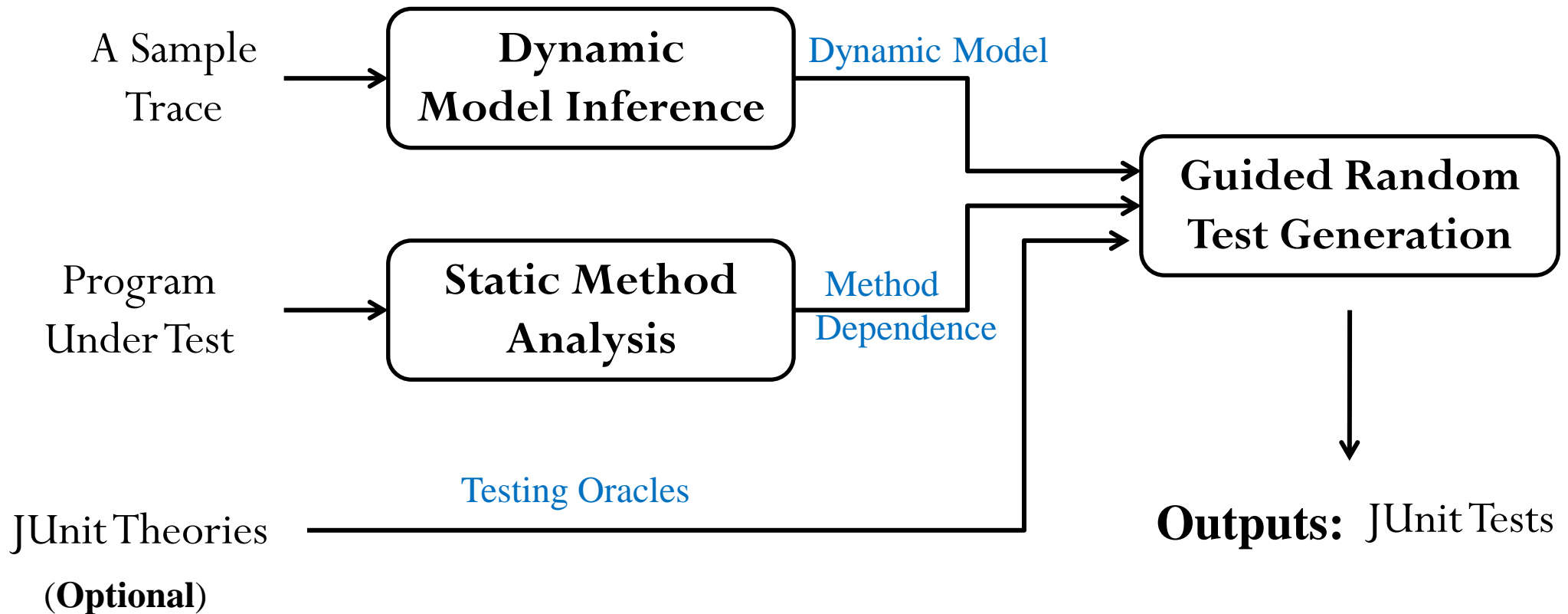


Outline

- Motivation
- Approach
 - Dynamic model inference
 - Static model expansion
 - Model-guided test generation
- Evaluation
- Related Work
- Conclusion and Future Work

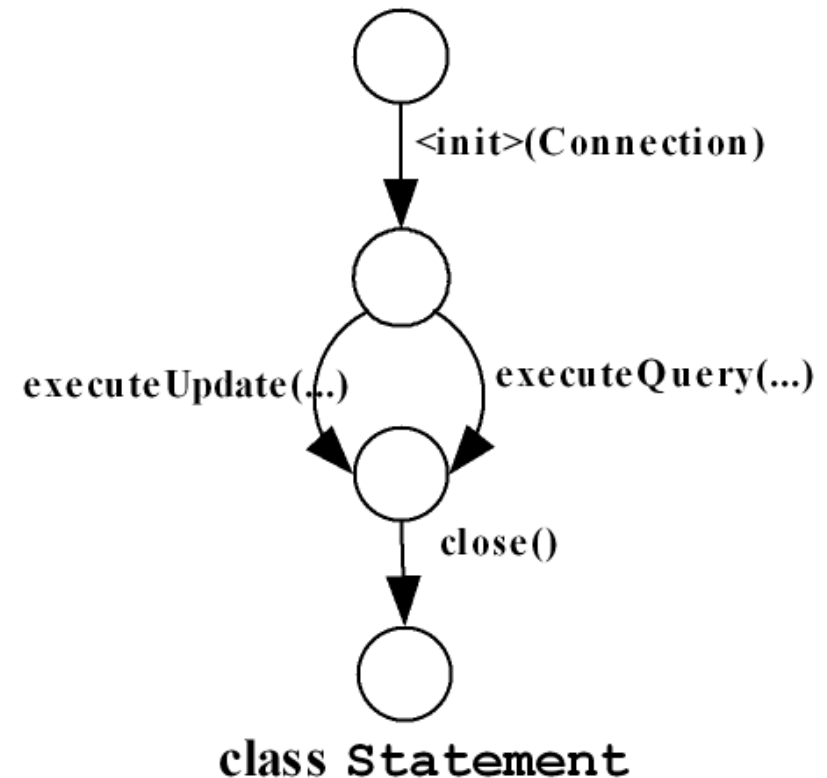
Overview of the Palus approach

Inputs:



(1) Dynamic Model Inference

- Infer a *call sequence model* for each tested class
 - Capture **possible ways** to create legal sequences
- A *call sequence model*
 - A **rooted, acyclic** graph
 - **Node**: object state
 - **Edge**: method-call
 - One model **per class**



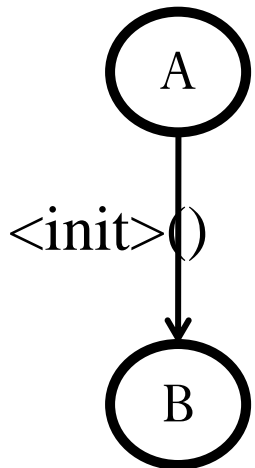
An Example Trace for Model Inference

```
Driver d = new Driver()  
Connection con = driver.connection("jdbc:dbname");  
  
Statement stmt1 = new Statement(con);  
stmt1.executeQuery("select * from table_name");  
stmt1.close();  
  
Statement stmt2 = new Statement(con);  
stmt2.executeUpdate("drop table table_name");  
stmt2.close();  
  
con.close();
```


Model Inference for class Driver

```
Driver d = new Driver();
```

Driver class

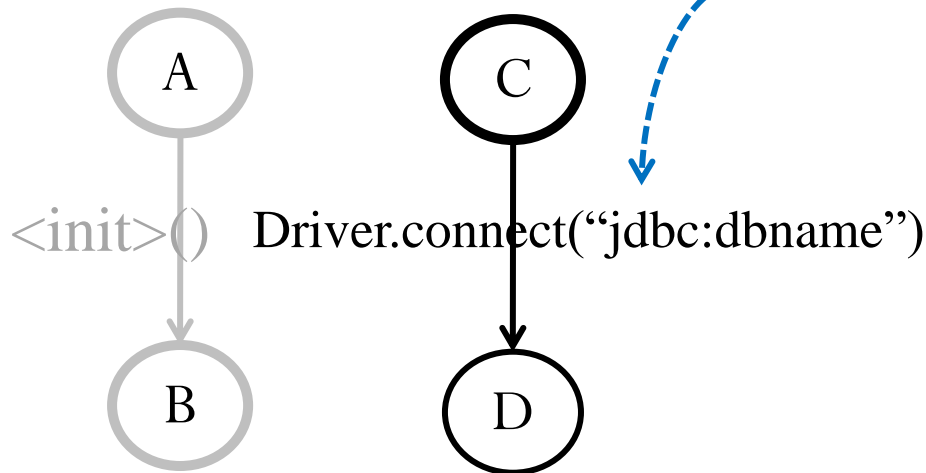


Model Inference for class Connection

```
Connection con = driver.connect("jdbc:dbname");
```

Driver class

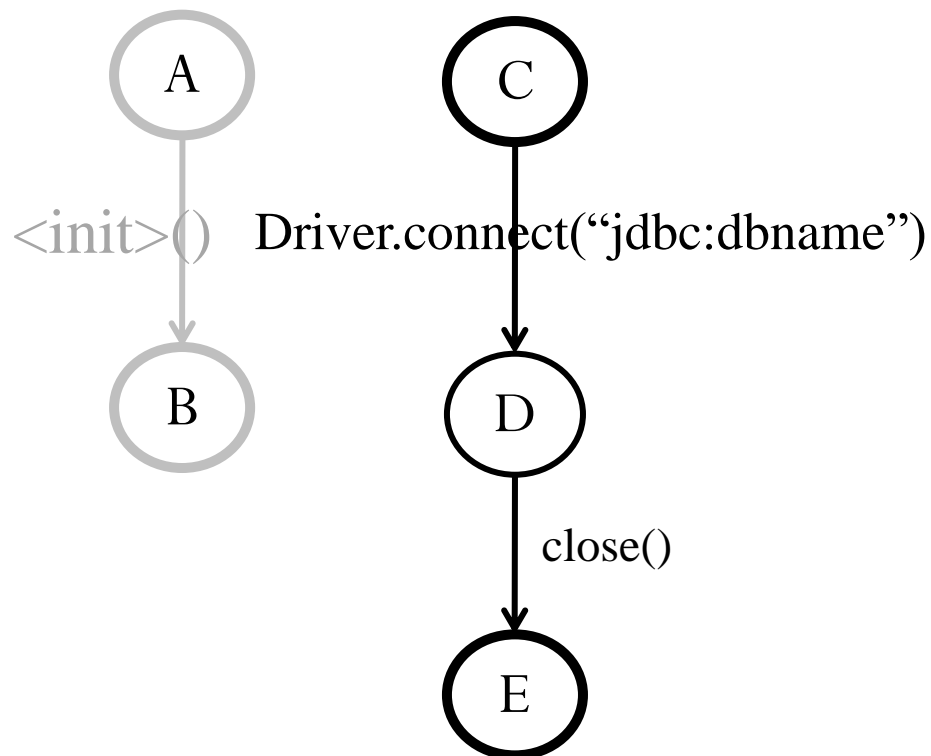
Connection class



Model Inference for class Connection

```
Connection con = driver.connect("jdbc:dbname");  
con.close();
```

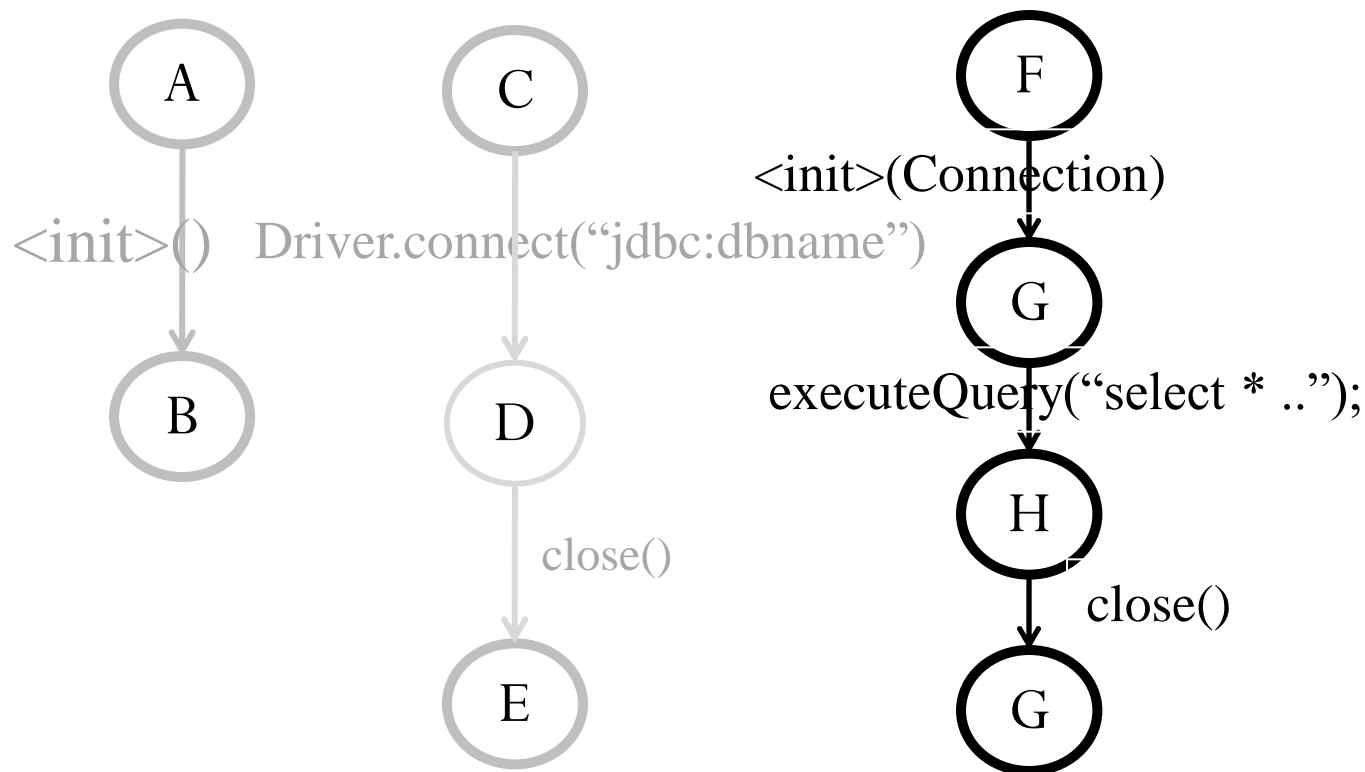
Driver class **Connection** class



Model Inference for class Statement

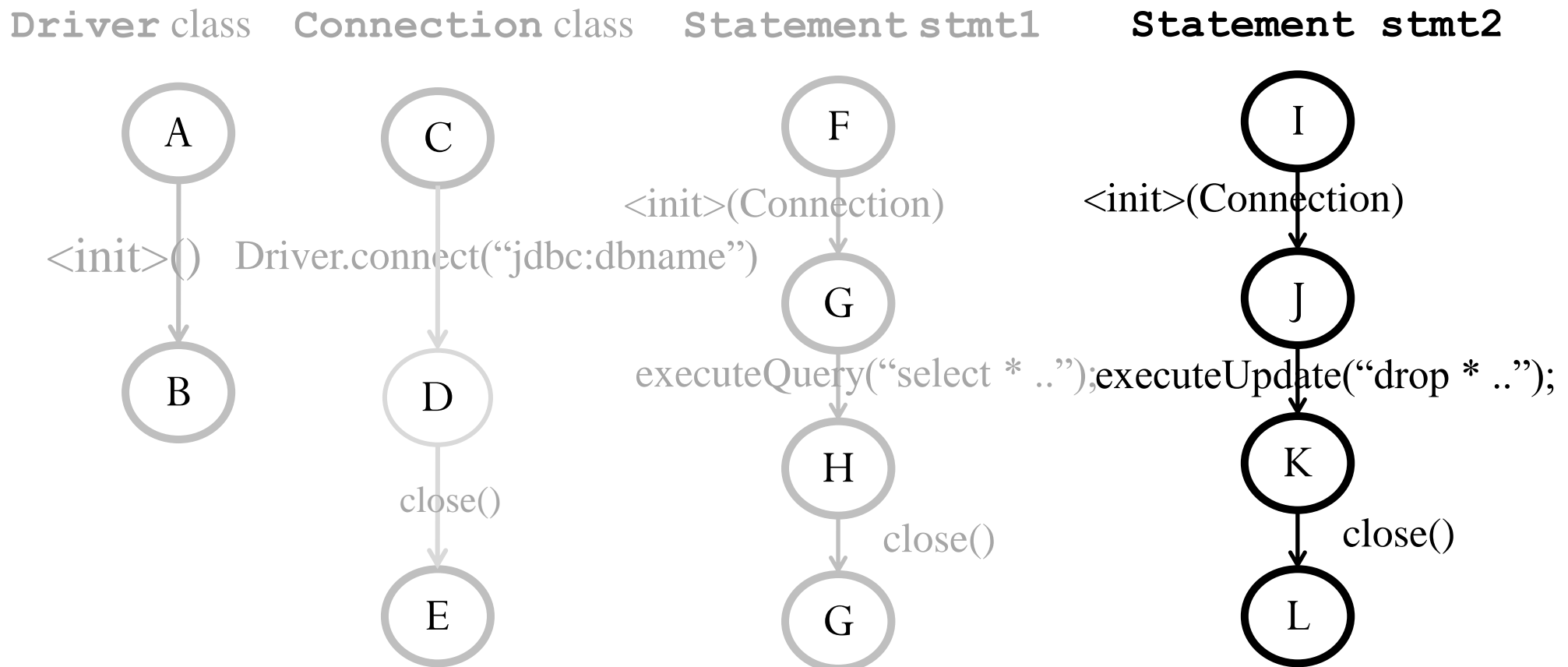
```
Statement stmt1 = new Statement(con);  
stmt1.executeQuery("select * from table_name");  
stmt1.close();
```

Driver class Connection class Statement stmt1

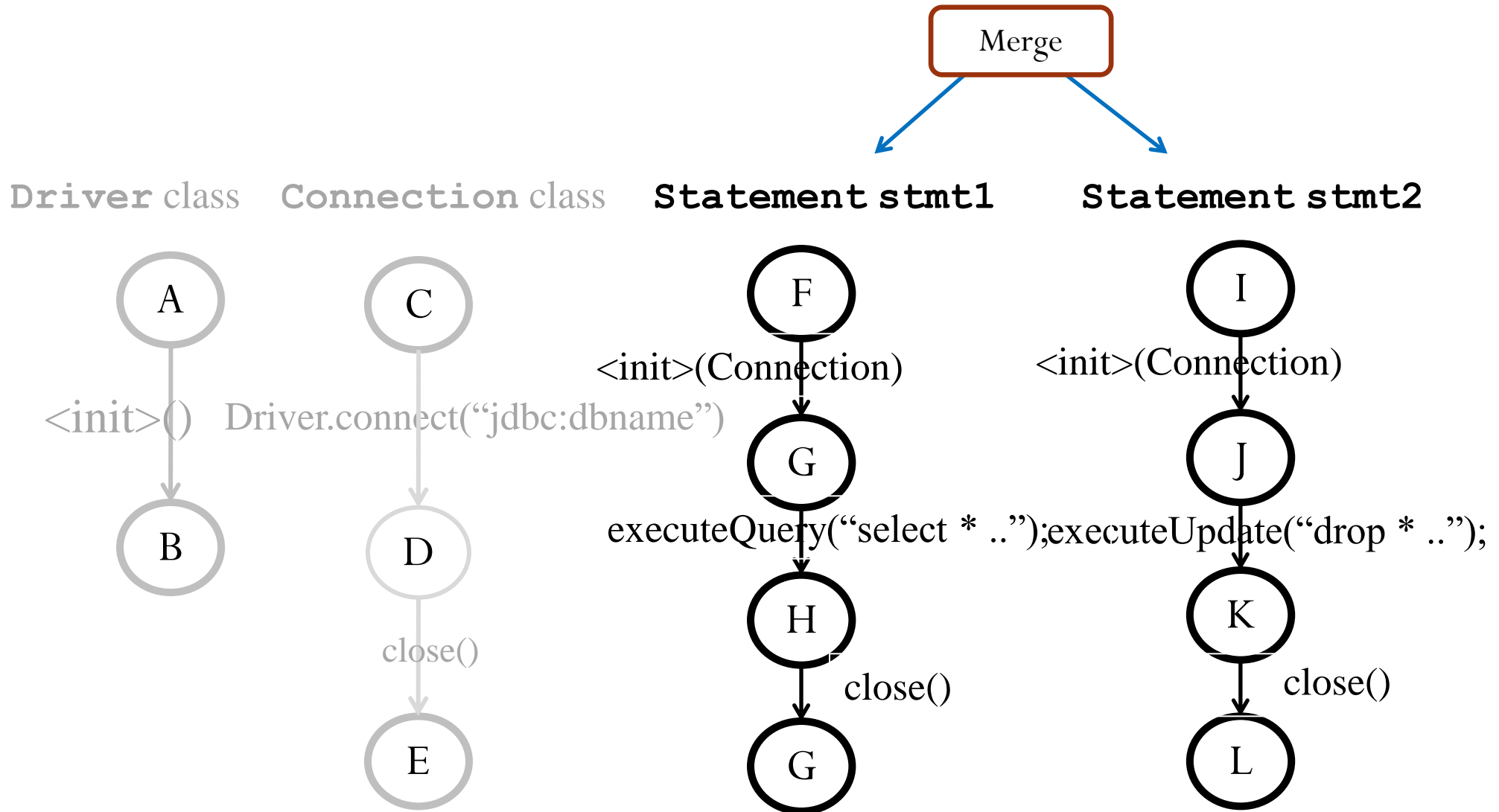


Model Inference for class Statement

```
Statement stmt2 = new Statement(con);  
stmt2.executeUpdate("drop table table_name");  
stmt2.close();
```



Merge Models of the Same class

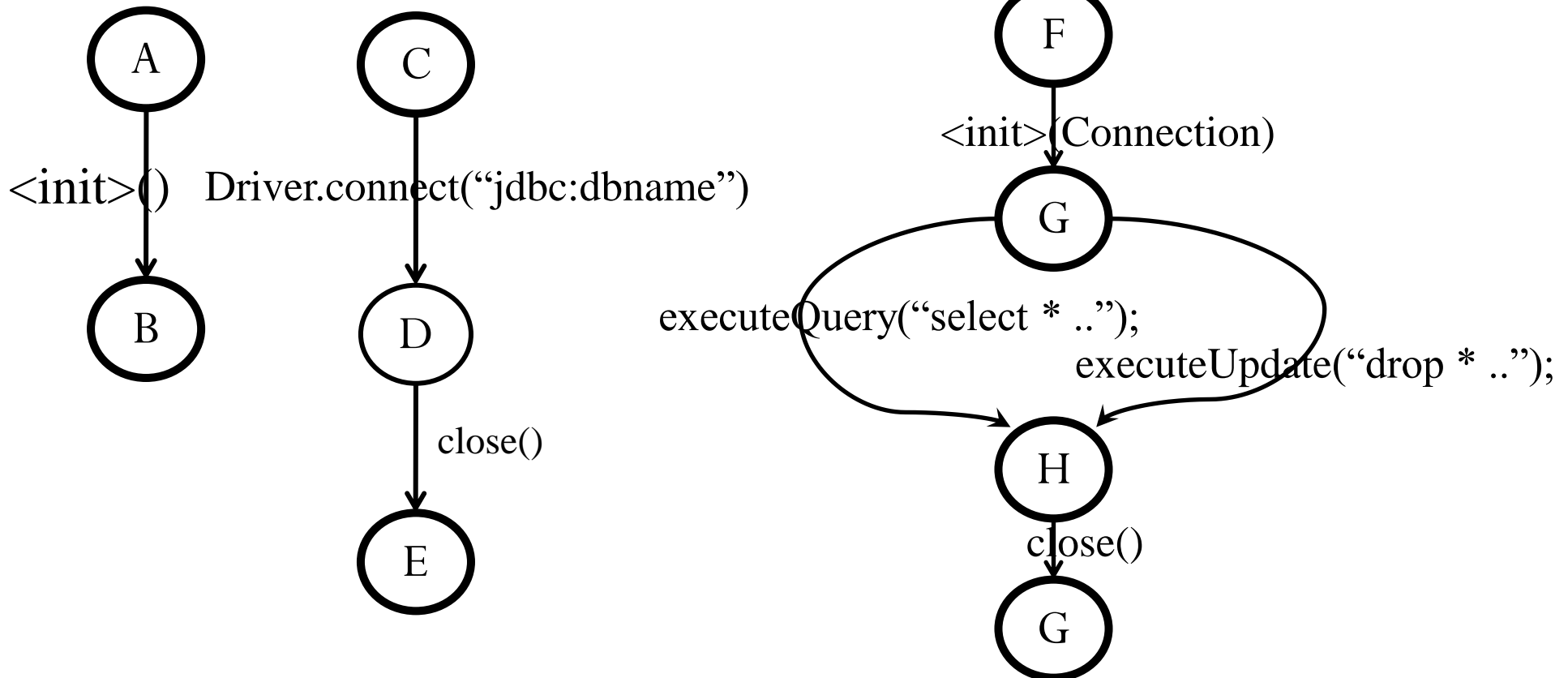


Call Sequence Model after Merging

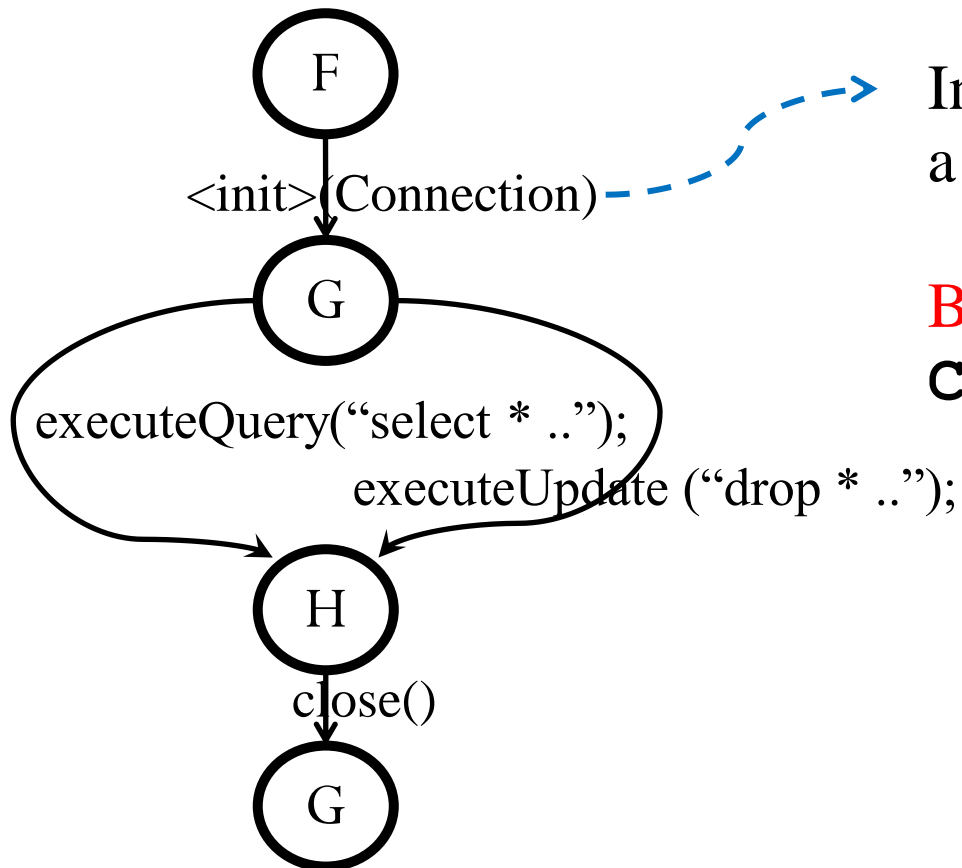
Driver class

Connection class

Statement class



Enhance Call Sequence Models with Argument Constraints



Invoking the constructor requires a **Connection** object

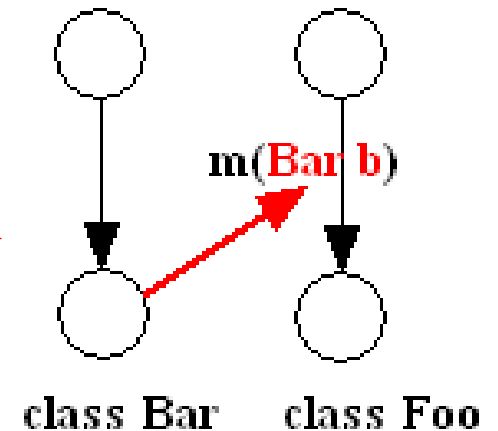
But, how to choose a desirable **Connection** object ?

Statement class

Argument Constraints

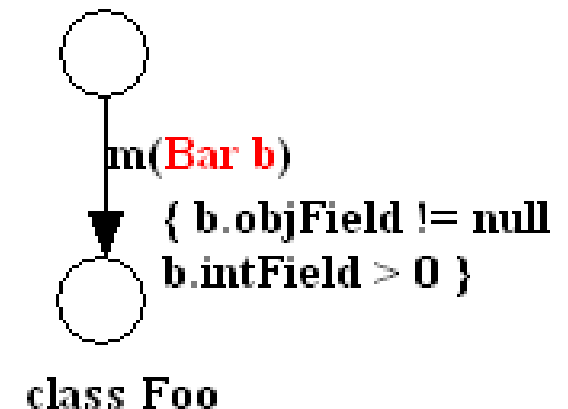
- **Argument dependence constraint**

- Record where the argument object values **come from**
- Add dependence edges in the call sequence models



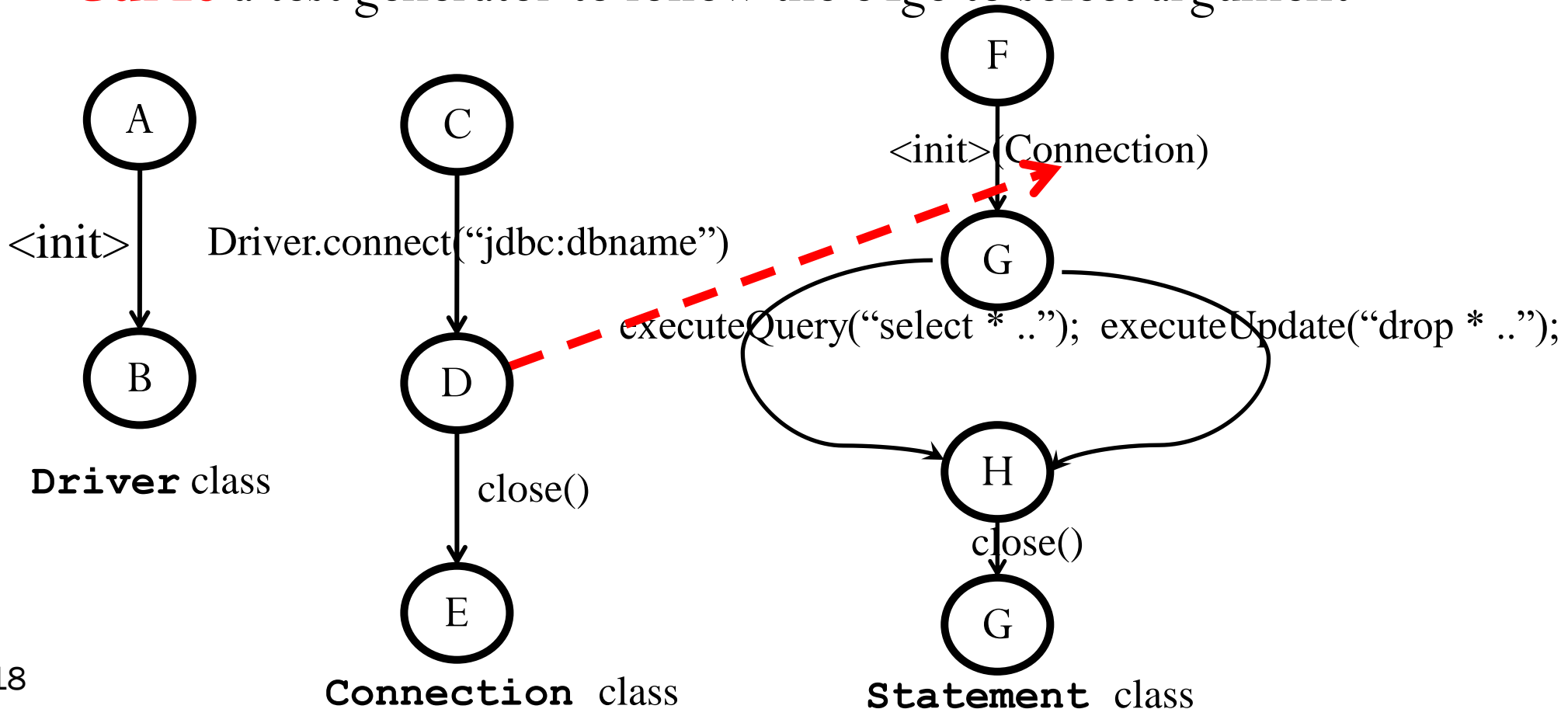
- **Abstract object profile constraint**

- Record what the argument **value “is”**
- Map each object field into an abstract domain
as a coarse-grained measurement of “value similarity”



Argument Dependence Constraint

- Represent by a directed edge ($- \rightarrow$ below)
- **Means:** transition $F \rightarrow G$ has data dependence on node D , it uses the result object at the node D
- **Guide** a test generator to follow the edge to select argument



Abstract Object Profile Constraint

- For each field in an observed object
 - Map the **concrete value** → an **abstract state**
 - Numeric value → $> 0, = 0, < 0$
 - Object → $= \text{null}, \neq \text{null}$
 - Array → $\text{empty}, \text{null}, \text{not_empty}$
 - Bool / enum values → not abstracted
- Annotate model edges with **abstract object profiles of the observed argument values** from dynamic analysis
- **Guide** test generator to choose arguments similar to what was seen at runtime

Annotate Model Edges with Abstract Object Profiles

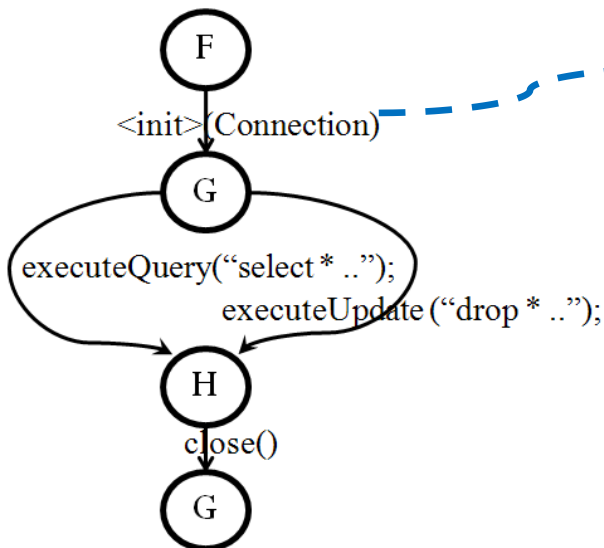
- Class **Connection** contains 3 fields

```
Driver driver; String url; String usr;
```

- All observed valid **Connection** objects have a profile like:

```
{driver != null, url != null, usr != null}
```

- Annotate the method-call edge: `<init>(Connection)`



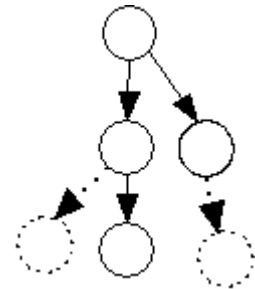
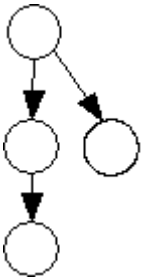
➤ Argument **Connection**'s profile:

```
{driver != null, url != null, usr !=null}
```

Palus prefers to pick an argument with the **same profile**, when invoking : `<init>(Connection)`

(2) Static Method Analysis

- Dynamic analysis is accurate, but **incomplete**
 - May fail to cover some methods or method invocation orders
- Palus uses static analysis to **expand** the dynamically-inferred model
 - Identify related methods, and test them together
 - Test methods not covered by the sample trace



Statically Identify Related Methods

- Two methods that access the same fields may be related (conservative)
- Two relations:
 - **Write-read**: method A reads a field that method B writes
 - **Read-read**: methods A and B reference the same field

Statically Recommends Related Methods for Testing

- Reach more program states
 - Call **setX()** before calling **getX()**
- Make the sequence more behaviorally-diverse
 - A correct execution observed by dynamic analysis will never contain:

```
Statement.close();  
Statement.executeQuery("...")
```
 - But static analysis may suggest to call **close()** before **executeQuery("...")**

Weighting Pair-wise Method Dependence

- **tf-idf weighting scheme** [Jones, 1972]
 - Palus uses it to measure the **importance** of a **field** to a **method**

$$tfidf(\mathit{field}, \mathit{method}) \propto \frac{\text{frequency of } \mathit{field} \text{ in } \mathit{method}}{\text{frequency of } \mathit{field} \text{ in all methods}}$$

- Dependence weight between two methods:

$$Weight(m1, m2) = \sum_{f \in \text{OverlapFields}(m1, m2)} tfidf(f, m1)$$

(3) Model-Guided Random Test Generation: A 2-Phase algorithm

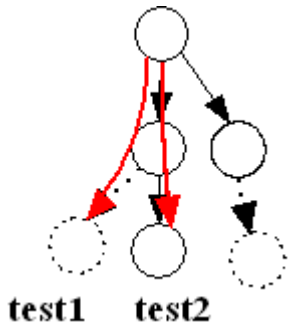
- Phase 1:

Loop:

1. Follow the **dynamically-inferred model** to select **methods** to invoke
2. For each selected **method**
 - 2.1 Choose arguments using:
 - Argument dependent edge
 - Captured abstract object profiles
 - Random selection
 - 2.2 Use **static method dependence** information to invoke related methods

- Phase 2:

- Randomly generate sequences for **model-uncovered** methods
- Use **feedback-directed** random test generation [ICSE'07]



Specify Testing Oracles in JUnit Theory

- A project-specific testing oracle in JUnit theory

```
@Theory
```

```
public void checkIterNoException(Iterator it) {  
    assumeNotNull(it);  
    try {  
        it.hasNext();  
    } catch (Exception e) {  
        fail("hasNext() should never throw exception!");  
    }  
}
```

Palus checks that, for **every** `Iterator` object, calling `hasNext()` should **never** throw exception!

Outline

- Motivation
- Approach
 - Dynamic model inference
 - Static model expansion
 - Model-guided test generation
- Evaluation
- Related Work
- Conclusion and Future Work

Research Questions

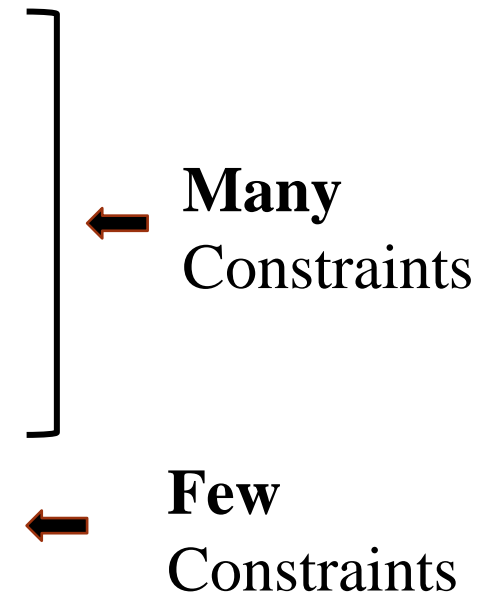
- Can tests generated by Palus achieve **higher structural coverage**
- Can Palus find (more) **real-world bugs**?
- Compare with three existing approaches:

Approaches	Dynamic	Static	Random
Randoop [ICSE'07]			●
Palulu [M-TOOS'06]	●		●
RecGen [ASE' 10]		●	●
Palus (Our approach)	●	●	●

Subjects in Evaluating Test Coverage

- 6 open-source projects

Program	Lines of Code
tinySQL	7,672
SAT4J	9,565
JSAP	4,890
Rhino	43,584
BCEL	24,465
Apache Commons	55,400



Experimental Procedure

- Obtain a sample execution trace by running a **simple example** from user manual, or its **regression test suite**
- Run each tool for until test coverage becomes **saturated**, using the **same trace**
- Compare the **line/branch coverage** of generated tests

Test Coverage Results

Approaches	Dynamic	Static	Random	Avg Coverage
Randoop [ICSE'07]			●	39%
Palulu [M-TOOS'06]	●		●	41%
RecGen [ASE' 10]		●	●	30%
Palus (Our approach)	●	●	●	53%

- Palus **increases** test coverage
 - Dynamic analysis helps to create **legal** tests
 - Static analysis / random testing helps to create **behaviorally-diverse** tests
- Palus **falls back** to **pure random** approach for programs with few constraints (Apache Commons)

Evaluating Bug-finding Ability

- Subjects:
 - The same 6 open-source projects
 - 4 large-scale Google products
- Procedure:
 - Check 5 **default Java contracts** for all subjects
 - Write 5 **simple theories** as additional testing oracles for Apache Commons, which has partial spec

Finding Bugs in 6 open-source Projects

- Checking default Java language contracts:
 - E.g., for a non-null object `o`: `o.equals(o)` returns true

	Dynamic	Static	Random	Bugs
Randoop [ICSE'07]			●	80
Palulu [M-TOOS'06]	●		●	76
RecGen [ASE' 10]		●	●	42
Palus (Our approach)	●	●	●	80

- Finds **the same number** of bugs as Randoop
- Writing additional theories as testing oracle
 - Palus finds **one new bug** in Apache Commons
 - `FilterListIterator.hasNext()` throws exception
 - **Confirmed** by Apache Commons developers

Finding Bugs in 4 Google Products

- 4 large-scale Google products

Google Product	Number of tested classes
Product A	238
Product B	600
Product C	1,269
Product D	1,455

- Each has a regression test suite with 60%+ coverage
- Go through a rigorous peer-review process

Palus Finds More Bugs

- Palus finds 22 real, previously-unknown bugs

	Dynamic	Static	Random	Bugs
Randoop [ICSE'07]			●	19
Palulu [M-TOOS'06]	●		●	18
RecGen [ASE' 10]		●	●	--
Palus (Our approach)	●	●	●	22

- **3 more** than existing approaches
- Primary reasons:
 - Fuzz a long specific *legal* path
 - Create a *legal* test, *diversify* it, and reach program states that have not been reached before

Outline

- Motivation
- Approach
 - Dynamic model inference
 - Static model expansion
 - Model-guided test generation
- Evaluation
- Related Work
- Conclusion and Future Work

Related Work

- Automated Test Generation
 - **Random approaches: Randoop [ICSE'07], Palulu [M-Toos'06], RecGen[ASE'10]**
Challenge in creating legal / behaviorally-diverse tests
 - **Systematic approaches: Korat [ISSTA'02], Symbolic-execution-based approaches (e.g., JPF, CUTE, DART, KLEE...)**
Scalability issues; create test inputs, not object-oriented method sequences
 - **Capture-replay -based approaches: OCAT [ISSTA'10], Test Factoring [ASE'05] and Carving [FSE'05]**
Save object states in memory, not create method sequences
- Software Behavior Model Inference
 - **Daikon [ICSE'99], ADABU [WODA'06], GK-Tail [ICSE'08] ...**
For program understanding, not for test generation

Outline

- Motivation
- Approach
 - Dynamic model inference
 - Static model expansion
 - Model-guided test generation
- Evaluation
- Related Work
- Conclusion and Future Work

Future Work

- Investigate **alternative ways** to use program analysis techniques for test generation
 - How to *better* combine static/dynamic analysis?
- What is a **good abstraction** for automated test generation tools?
 - We use an enhanced call sequence model in Palus, what about other models?
- Explain **why a test fails**
 - Automated Documentation Inference [ASE'11 to appear]
 - Semantic test simplification

Contributions

- A hybrid automated test generation **technique**
 - **Dynamic analysis**: infer model to create legal tests
 - **Static analysis**: expand dynamically-inferred model
 - **Random testing**: create behaviorally-diverse tests
- A publicly-available **tool**
<http://code.google.com/p/tpalus/>
- An empirical **evaluation** to show its effectiveness
 - Increases test coverage
 - Finds more bugs

Backup slides

Sensitivity to the Inputs

- Investigate on two subjects: **tinySQL** and **SAT4J**

Subject	Input Size	Coverage
tinySQL	10 SQL Statements	59%
	ALL Statements from Manual	61%
SAT4J	A 5-clause formula	65%
	A 188-clause formula	66%
	A 800-clause formula	66%

- This approach is **not very sensitive** to the inputs
 - Not too many constraints in subjects?

Breakdown of Contributions in Coverage Increase

