# **Scaling Java Points-To** Analysis Using **SPARK** (Soot Pointer Analysis Research Kit) Ondřej Lhoták and Laurie Hendren Sable Research Group **McGill University** April 8th, 2003

#### **Problems**

- Implementing a points-to analysis to handle the details of Java
  - is a lot of work.
  - is difficult to do correctly.
- Research done on disparate implementations is often incomparable.

# **Objectives**

- Develop a flexible, efficient framework for experimenting with variations in Java points-to analyses
- Demonstrate its usefulness with an empirical comparison of precision and efficiency of some of these variations

# Outline

- Spark overview
- Empirical study
- Overall performance
- Uses of Spark
- Conclusion

#### **Spark overview**

- Part of Soot bytecode transformation and annotation framework [CC 00] [CC 01]
- Initial representation is Soot's Jimple
  - Typed [SAS 00]
  - Three-address (only simple operations)
- Spark internal representation is Pointer Assignment Graph (PAG)
  - Nodes for variables, allocation sites, field references
  - Edges representing subset constraints

#### **Spark overview**



- Analysis variations expressed by building different PAGs for the same code
- This talk concentrates on flow-insensitive, subset-based variations

## **Empirical study**

# Factors affecting precision

- Enforcing declared types
- Field reference representation
- Call graph construction
- Factors affecting only efficiency
  - Pointer assignment graph simplification
  - Set implementation
  - Propagation algorithms

#### **Declared types: ignore**



#### **Declared types: ignore**



#### **Declared types: ignore**





#### **Declared types: enforce during analysis**



# **Enforcing declared types**

- ignoring types produces many large sets (> 1000 elements) of spurious points-to relationships
- in practice, enforcing types after analysis almost as precise as during analysis
- enforcing types during analysis prevents blowup during the analysis

ignore	slow	less precise
after analysis	slow	more precise
during analysis	fast	more precise

#### **Empirical study**

# Factors affecting precision

- Enforcing declared types
- Field reference representation
- Call graph construction
- Factors affecting only efficiency
  - Pointer assignment graph simplification
  - Set implementation
  - Propagation algorithms

#### **Field representation**

- Field references can be represented in different ways:
  - field-sensitive distinguishes fields of different objects
  - field-based ignores the base object, grouping all objects having the field together

#### **Field-sensitive representation**



#### **Field-based representation**

#### **Field representation**

- Field-sensitive requires iterating
- Field-based less precise, but possible in a single iteration
- Clever propagation algorithm can make speed difference very small

field-based	very fast	less precise
field-sensitive	almost as fast	more precise

## **Empirical study**

# Factors affecting precision

- Enforcing declared types
- Field reference representation
- Call graph construction
- Factors affecting only efficiency
  - Pointer assignment graph simplification
  - Set implementation
  - Propagation algorithms

# **Call graph construction**

- An approximation of the call graph is required for points-to analysis
- It can be built
  - ahead-of-time using an analysis such as Class Hierarchy Analysis
  - on-the-fly during the analysis as actual types of receivers are computed

#### Call graph construction: CHA



#### Call graph construction: on-the-fly



#### Call graph construction: on-the-fly

Hierarchy B.foo() this Х return B class B { foo() { . . . } } C.foo() class C { foo() { . . . } } returr A x = new B();A y = x.foo();

this

# **Call graph construction**

- Building call graph on-the-fly requires adding edges during propagation
  - requires more iteration
  - reduces simplification opportunities before propagation
- CHA call graph includes more spurious, unreachable methods than on-the-fly

CHA	fast	less precise
on-the-fly	slow	more precise

# **Empirical study**

# Factors affecting precision

- Enforcing declared types
- Field reference representation
- Call graph construction
- Factors affecting only efficiency
  - Pointer assignment graph simplification
  - Set implementation
  - Propagation algorithms

- Groups of nodes can be merged [Rountev,Chandra 00]
  - strongly-connected components
  - single-entry subgraphs



- Groups of nodes can be merged [Rountev,Chandra 00]
  - strongly-connected components
  - single-entry subgraphs



- Groups of nodes can be merged [Rountev,Chandra 00]
  - strongly-connected components
  - single-entry subgraphs



- Groups of nodes can be merged [Rountev,Chandra 00]
  - strongly-connected components
  - single-entry subgraphs



Factors limiting simplification opportunities

- Enforcing declared types changes points-to sets
- On-the-fly call graph eliminates edges from initial pointer assignment graph



# **Empirical study**

## Factors affecting precision

- Enforcing declared types
- Field reference representation
- Call graph construction

# Factors affecting only efficiency

- Pointer assignment graph simplification
- Set implementation
- Propagation algorithms

#### **Set implementation**

- hash Using java.util.HashSet
- array Sorted array, binary search



bit Bit vector

#### hybrid

- Array for small sets
- Bit vector for large sets

## **Set implementation**

hash	slow	large
array	slow	small
bit	fast	large
hybrid	fast	small

In the above table,

- slow is up to 100 times slower than fast
- Iarge is up to 3 times larger than small
- Set implementation is very important

# **Empirical study**

## Factors affecting precision

- Enforcing declared types
- Field reference representation
- Call graph construction

# Factors affecting only efficiency

- Pointer assignment graph simplification
- Set implementation
- Propagation algorithms

#### **Propagation algorithms: iterative**

repeat for each edge *e* propagate along *e*; end for until no change

Slightly more complicated to handle

- field references
- on-the-fly call graph

#### **Propagation algorithms: worklist**

while worklist not empty do
 remove node n from worklist;
 for each edge e starting at n
 propagate along e;
 add all affected nodes to worklist;
 end for
end while

#### **Propagation algorithms: worklist**

while worklist not empty do
 remove node n from worklist;
 for each edge e starting at n
 propagate along e;
 add all affected nodes to worklist;
 end for
end while

- With field references, difficult to determine affected nodes
- Very costly to determine all affected nodes due to of aliasing

#### **Propagation algorithms: worklist**

## repeat while worklist not empty do remove node *n* from worklist; for each edge e starting at n propagate along e; add most affected nodes to worklist; end for end while propagate along all field reference edges; until no change

Solution: find most affected nodes, and add outer loop to handle missed nodes





Ist iteration: propagate {A,B,C,D}





- Ist iteration: propagate {A,B,C,D}
- add E to x
- 2nd iteration: propagate {A,B,C,D,E}



Idea: split sets into new and old part



Ist iteration: propagate {A,B,C,D}



- Ist iteration: propagate {A,B,C,D}
- flush new to old



- Ist iteration: propagate {A,B,C,D}
- flush new to old
- add E to x



- Ist iteration: propagate {A,B,C,D}
- flush new to old
- add E to x
- 2nd iteration: propagate {E}



- Ist iteration: propagate {A,B,C,D}
- flush new to old
- add E to x
- 2nd iteration: propagate {E}
- flush new to old

#### **Propagation algorithms**

- When to use worklist?
  - Always, about twice as fast as iterative
- When to use incremental worklist?
  - Always, except with CHA call graph field-based analysis, in which there is not enough iteration

# **Summary of findings**

- Declared types should be enforced during propagation for a scalable analysis
- Hybrid set implementation much faster than others, up to 2 orders of magnitude, with reasonable memory consumption
- Field-based can be done in one iteration, but field-sensitive with worklist algorithm is almost as fast and slightly more precise
- Tradeoff: On-the-fly call graph slower but more precise than ahead-of-time CHA call graph

# Outline

- Spark overview
- Empirical study
- Overall performance
- Uses of Spark
- Conclusion

- Rountev, Milanova, Ryder [OOPSLA 01]
  - 360 MHz SPARC, solver written in ML
  - version 1.1.8 library (150 KLOC)
- Whaley, Lam [SAS 02]
  - 2 GHz Pentium, solver written in Java
  - version 1.3.1 library (500 KLOC)
  - optimistic call graph (potentially unsafe)
- (Spark) Lhoták, Hendren [CC 03]
  - 1.67 GHz Athlon, solver written in Java
  - version 1.3.1 library (500 KLOC)

Common metric: number of methods analyzed

#### **Overall performance: time**



#### **Overall performance: space**



# Outline

- Spark overview
- Empirical study
- Overall performance
- Uses of Spark
- Conclusion

## **Uses of Spark**

- Use points-to and side-effect information in Soot analyses
- Encode in attributes
  - for use in JITs
  - for use in program understanding
- Experiment with points-to algorithms
  - using Spark command-line switches
  - by implementing new algorithms within Spark

#### Conclusions

- Spark is a flexible and efficient framework for experimenting with variations in Java points-to analyses
- We have demonstrated its usefulness in an empirical study of some of these variations

# Ongoing work

- BDD-based solvers [PLDI 03]
- Object-sensitivity [Milanova,Rountev,Ryder 02]
- On-the-fly cycle detection [Heintze, Tardieu 01]
- Shared bit-vector [Heintze, Tardieu 01]

# **Obtaining Spark**

- Spark is part of Soot since version 1.2.4
- Soot is available under the LGPL
  - http://www.sable.mcgill.ca/soot
- Future plans for Soot
  - Major update (version 2.0) in June 2003
  - Tutorial at PLDI