Demand-Driven Points-To Analysis For Java

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Who needs better pointer analysis?

IDEs:

- for <u>refactoring</u>, program understanding
- but program edits <u>invalidate</u> analysis
- current analyses too slow to re-analyze
- incremental analyses hard to engineer

JIT compilers:

- for virtual call resolution, register allocation
- but current analyses too slow for runtime
- plus, class loading invalidates, needs re-analysis

What we provide

Analysis so <u>fast</u> that you can

- run it in the JIT compiler
 - 16x speedup compared to Andersen's analysis
- rerun it after the code changes
 - 2ms per query about a pointer variable

Analysis with <u>low memory overhead</u>

< 50 KB, eases engineering effort

Analysis with tunable precision

adjustable to different time constraints

Contributions

- 1) Demand analysis with early termination:
 - return conservative result after a time out

Problem we had to solve:

- how to <u>approximate</u> to make early termination rare?
- 2) Refining the approximation

Problems we had to solve:

- Mechanism: how to refine?
- Policy: where to focus the refinement budget?

Outline

- Points-to analysis background
- Our approach
 - Demand analysis
 - Early termination
- Our algorithms
 - CFL-reachability formulation
 - Approximation
 - Refinement (undoing the approximation)
- Experiments

Points-To Analysis

- Compute objects each variable can point to
 - For each var x, points-to set pt(x)
- Andersen's Analysis: our reference point
 - Want similar precision for our analysis
 - One <u>abstract location</u> for each allocation site
 x = new Foo() yields pt(x) = { o1_{Foo} }
 - Context- and flow-insensitive
- Current implementations not suitable for us
 - Too costly for JIT, IDE
 - 30 s / 30 MB (Berndl et. al. PLDI03) on jedit
 - Code changes require re-analysis

Demand-Driven Analysis

Protocol:

- Client asks a <u>query</u>: what's the points-to set of variable x?
- Analysis computes only the points-to set of x

Works well when typically few queries:

- JIT compiler: variables in hot code
- IDE: variables in code being edited by developer

Visits <u>theoretically minimal</u> set of statements Problem:

- worst-case time same as exhaustive
- Happens in practice for standard Andersen's
- Lesson: Need to <u>approximate</u> for scalability
 - Ideally, maintain nearly all precision

Approx: Early Termination

Terminate queries when budget exhausted Return a sound result to client

- early result: pt(x) = { all abstract locs }
- complete result: $pt(x) = \{ o1_{Foo}, o2_{Bar} \}$

No precision loss if complete result does not satisfy client

Hypothesis: long-running) unsatisfying

- Suggested previously (Heintze / Tardieu PLDI01)
- For standard Andersen's, large precision loss

Challenge: how to approximate further?

Key Ideas

Formulate analysis in CFL-reachability

- Natural for demand-driven analysis
- Andersen's for Java is <u>balanced parens</u>

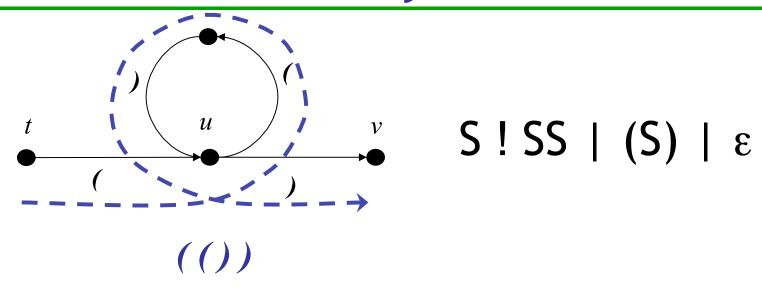
Approximate through regularization

Solvable by linear DFS algorithm

<u>Iterative refinement</u> to de-approximate

- Simple recursive queries
- Client-driven

CFL-Reachability



Points-to analysis graph:

- Nodes represent variables / locs
- Edges represent statements

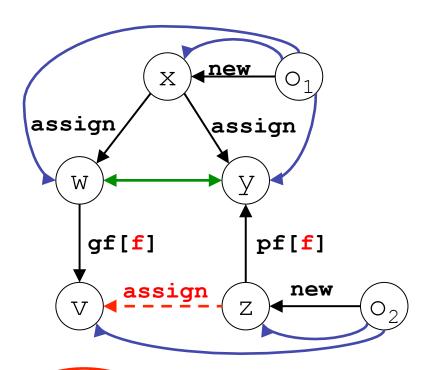
Points-to analysis paths:

- o 2 pt(x), *flowsTo*-path from o to x
- pt(x) Å pt(y) ≠ ; , alias-path from x to y

Andersen's Analysis in CFL-Reachability

```
x = \text{new Obj}(); // o_1
z = \text{new Obj}(); // o_2
w = x:
y = x;
y.f = z;
v = w.f;
```

Edge types statement flowsTo alias



flows to ! new (psssipp) kias of [f]





assign) *

balanced parens

Field-sensitive formulation: standard for Java

See paper for alias grammar

Approx: Regularization

Add match edges for matching

field read/write pairs

- From source of putfield
- To sink of getfield

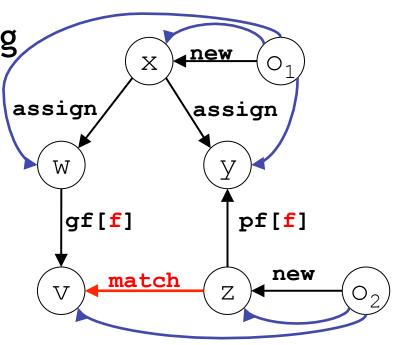
Regular grammar

Yields DFS algorithm

Field-based precision

flowsToRegnewner [(finatachias agssingh) * assign) *

```
pf[f] alias gf[f] ) match
o flowsTo x ) o flowsToReg x
```



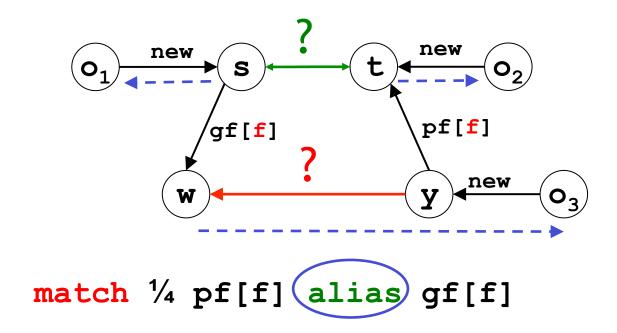
RegularPT

```
marked, worklist: Set of Node

procedure query(source: Node)
add source to marked and worklist
while (worklist is non-empty) do
remove w from worklist
foreach NEW edge o -> w do
add o to points-to set of source
end
foreach ASSIGN and MATCH edge y -> w do
if y unmarked, add y to marked and worklist
end
end
end
```

- No caching, so very low memory usage
- Early termination through traversal budget

Refining Match Edges



Most approximation can be refined

• Imprecise for recursive fields

Client-Driven Refinement Policy

Not clear when / where to refine

- Extra queries may be costly
- Refining match edge may not affect result

Client-driven: only refine when client affected

- E.g, multiple targets for virtual call
- Guyer and Lin SAS03

RefinedRegularPT:

- Refine edges traversed by RegularPT
- Iterate until client satisfied or budget exhausted

Experimental Hypotheses

- 1) Algorithms precise with early termination
 - Regular approximation reasonable
 - Refinement yields improved precision
- 2) Algorithms meet performance goals
 - Fast running time
 - Low memory

Evaluation Framework

Implemented in Soot / SPARK framework Benchmarks: SPEC, Ashes, jedit Clients

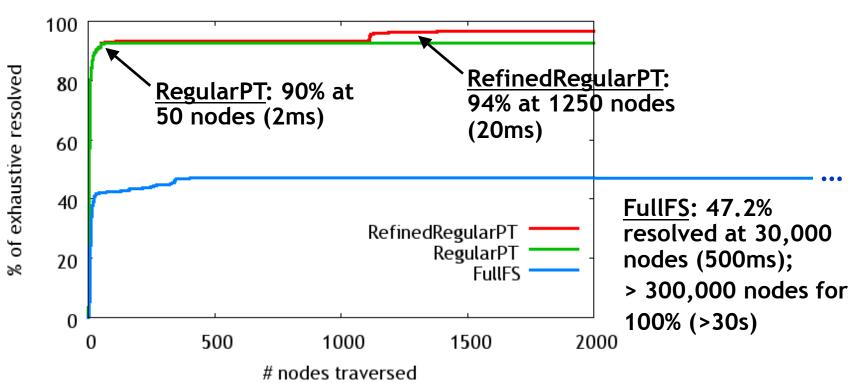
- IDE: Virtual call resolution
 - For program understanding
- JIT: queries from hot code
 - Virtual call resolution (for inlining)
 - Local aliasing (for load/store elimination)

Algorithms

| Name | Field-based / sensitive | Demand / Exhaustive | Notes |
|------------------|--|------------------------|--|
| RegularPT | Field-based | Demand | DFS Traversal |
| RefinedRegularPT | Variable up to partially field-sensitive | Demand | Client-driven refinement |
| FullFS | Field-sensitive | Demand | Heintze and Tardieu [PLDI01] adapted to Java |
| ExhaustiveFB | Field-based | Exhaustive | from SPARK |
| ExhaustiveFS | Field-sensitive | Exhaustive | from SPARK |

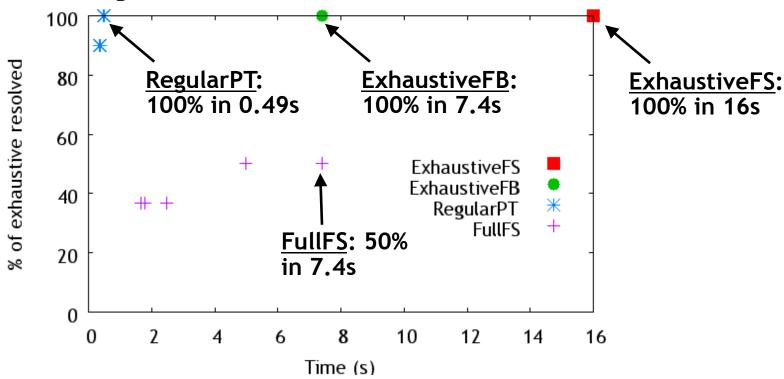
1) Evaluation: Precision





2) Evaluation: Performance





Memory:

- < <u>50 KB</u> for (Refined)RegularPT
- 28MB for FullFS using BDDs

Conclusions

New demand points-to analysis

- Speed through two approximations
 - Early termination
 - Regularization
- Refinement driven by client

Provide <u>high precision</u> in <u>tight budget</u> Suitable for JITs, IDEs; and elsewhere?