Two Birds with One Stone: Multi-Derivation for Fast Context-Free Language Reachability Analysis

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CFL (Context-Free Language) Reachability

- Fundamental framework for program analysis
 - Taint Analysis
 - Pointer Analysis
 - Bug Detection
 - Program Slicing
 - ...

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Edge-labeled Graph G

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X ::= a b Y ::= X c

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Context-free Grammar of L

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Edge-labeled Graph G

Context-free Grammar of L

CFL solving ~ Edge Derivation + Edge Insertion

An X-reachability relation holds between Node 0 and Node 2, i.e., Node 2 is X-reachable from Node 0



Edge-labeled Graph G

X ::= a b ← Y ::= X c

Context-free Grammar of L

Insert an X-edge from Node 0 to Node 2!



X ::= a b *Y* ::= *X c*

A Y-reachability relation holds between Node 0 and Node 3



X ::= *a b* Y ::= X c 🛏

Insert a Y-edge from Node 0 to Node 3!



X ::= *a b* Y ::= X c 🔶

Limitation of Existing Approaches





single-reachability derivation: existing CFL algorithms process multiple areachability relations *separately*, which causes *redundancy*





Single-reachability derivation

X ::= a b Y ::= X c

Packing multiple a-reachability relations together





Single-reachability derivation

X ::= a b ← Y ::= X c

Propagating a-reachability relations in **batch** via b-edge -> multiple new X-reachability relations produced



0 a b 2 c 3 n a 1 b 2 c 3

Single-reachability derivation

X ::= a b Y ::= X c ←

Propagating X-reachability relations in batch via c-edge -> multiple new Y-reachability relations produced



0 a b 2 c 3 n a

Single-reachability derivation

X ::= a b Y ::= X c

Difference propagation!

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$$1 \xrightarrow{A} 2 \xrightarrow{A} 3$$

Redundant Propagation due to Transitivity

The X-reachability relations of Node 1 are propagated **twice** from Node 1 to Node 3



A ::= A AX ::= X A

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Shortcut path causes redundant propagation!

- A <u>transitivity-aware</u> propagation graph, **PG(A)** for transitive relation A
 - With redundant edges excluded (partially)

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- A <u>transitivity-aware</u> propagation graph, **PG(A)** for transitive relation A
 - With redundant edges excluded (partially)
- Construction
 - Subgraph induced from the original edge-labeled graph
 - Constructed on the fly



- The construction of propagation graph is non-trivial
 - Duplicate A-edges can be introduced by other productions rather than A ::= A A
 - Can be seen as partial transitive reduction
 - Please refer to our paper for more details



Solving Transitivity via Multi-Derivation

- A ::= A A: propagating A-reachability relations in batch in PG(A)
- X ::= X A: propagating X-reachability relations in batch in PG(A)

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Solving Transitivity via Multi-Derivation

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- PG(A) is constructed on the fly with redundant edges excluded
- X ::= A X can be handled similarly as X ::= X A

Evaluation

- Implementation: SVF, <u>https://github.com/SVF-tools/SVF</u>
 - CFL-reachability solver: Pearl
- 2 popular clients: Value-Flow Analysis and Alias Analysis for C++
- 10 benchmarks from SPEC CPU2017 C/C++
- Compare with
 - the standard algorithm
 - POCR[OOPSLA'22] for fast transitivity solving

Performance

Field-Sensitive Alias Analysis

Speed up over POCR: 2.4x (avg.), 4.2x(max.)

id	Std		Pocr			PEARL		
	Time	Mem	Time	SPU	Mem	Time	SPU	Mem
cactus	-	-	191.27	-	11.62	96.59	2.0x	9.28
imagick		. 	554.13	-	42.55	334.76	1.7x	41.41
leela	312.28	0.31	3.40	91.8x	0.39	2.24	1.5x	0.36
nab	7.12	0.10	0.76	9.4x	0.10	0.18	4.2x	0.09
omnetpp	-		410.79	-	17.96	195.77	2.1x	17.08
parest	-		92.77	-	4.79	42.10	2.2x	4.69
perlbench	-	<u> 1</u> 20	1733.42	-	110.84	978.29	1.8x	80.30
povray	14699.10	3.24	160.97	91.3x	6.60	58.64	2.7x	5.55
x264	1056.20	1.31	11.13	94.9x	1.05	3.39	3.3x	1.00
XZ	6.67	0.05	0.42	15.9x	0.07	0.19	2.2x	0.07

Performance

Context-Sensitive Value Flow Analysis

Speed up over POCR: 10.1x (avg.), 29.2x(max.)

id	Std		Pocr			PEARL		
	Time	Mem	Time	SPU	Mem	Time	SPU	Mem
cactus	3408.36	3.46	604.10	5.6x	40.26	28.26	21.4x	4.74
imagick	583.71	0.43	59.13	9.9x	5.87	5.18	11.4x	0.74
leela	1.58	0.02	0.47	3.4x	0.19	0.16	2.9x	0.02
nab	55.51	0.50	16.59	3.3x	4.34	3.27	5.1x	0.32
omnetpp	229.26	1.08	15.49	14.8x	3.99	3.62	4.3x	0.53
parest	2.40	0.07	0.67	3.6x	0.19	0.38	1.8x	0.07
perlbench	16366.80	6.35	1520.19	10.8x	63.57	52.06	29.2x	10.42
povray	5834.13	5.05	655.14	8.9x	55.84	43.91	14.9x	4.72
x264	194.16	0.70	34.77	5.6x	6.46	4.71	7.4x	0.67
xz	0.54	0.01	0.16	3.4x	0.06	0.06	2.7x	0.01

Performance

Context-Sensitive Value Flow Analysis

For *perlbench*, reduce **84% memory usage** over POCR

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Ablation Study

- Pwb (Pearl without batch propagation)
- Value-Flow Analysis: Pwb is 7.2x faster than POCR, and 1.3x slower than Pearl
- Alias Analysis: Pwb is comparable with POCR, and 2.2x slower than Pearl

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- Reason of different speedups
 - Propagation graph is simple yet effective, especially when transitive relations dominates, e.g., in value-flow analysis.
 - Value-flow analysis has no productions in the form of X ::= X A, so the effectiveness of multi-derivation is limited.

Effectiveness of Multi-Derivation

How many propagations of transitive relations can be reduced?

Reduction rates (Pearl v.s. Pwb)

- Value-flow analysis: 79.3%
- Alias analysis: 98.5%



Reduction rates in propagations of transitive relations

Conclusion

- Contributions
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 - A highly efficient CFL-reachability solver, Pearl

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- Artifact available
 - Docker image and instructions for reproduction
 - https://figshare.com/articles/dataset/ASE_2023_artifact/23702271



Thank you for your listening!

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Backup Slides

Set Constraint

- Set constraint and CFL reachability are interconvertible
- Two steps
 - reduce the CFL problem to a set contraint instance
 - solve it using off-the-shelf constraint solver
- Disadvantages
 - Unawareness of the graph features, e.g., transitivity
 - An extra reduction step