Scaling CFL-Reachability-Based Alias Analysis: Theory and Practice

Qirun Zhang

Thesis Defense



Scaling CFL-Reachability-Based Alias Analysis: Theory and Practice

Context-free language reachability

4

Context-free language (CFL) reachability

- A directed graph G = (V, E)
- A context-free grammar $CFG = (\Sigma, N, P, S)$
- Each $(u, v) \in E$ is labeled with $L(u, v) \in \Sigma$
- Realized string R(p) of a path p

•
$$R(p) = \overline{a}_1 a_2 \overline{a}_2$$
, where $p = 2, 3, 4, 5$

S-path: R(p) can be generated from start symbol S
 Summary edge (u, S, v) and S-edge

Query

- \blacksquare Is there an S-path from u to v ?
- Single-source-single-sink

All-pairs



Scaling CFL-Reachability-Based Alias Analysis: Theory and Practice

Alias analysis

Alias analysis

- Determine whether two pointer variables contain the same memory location in any execution of the program.
- A fundamental static analysis problem.
- A client analysis
 - Buffer overflow

```
char *a, *b;
char c[10], d[11];
...
strcpy(a, b);
```

CFL-reachability-based alias analysis

7

Alias analysis via CFL-reachability

```
int x;
int *u, *w, *y, *z;
int **v;
u = *v;
*v = w;
v = &y;
y = &x;
z = y;
```



(b) The corresponding PEG.

(a) A code snippet.

Main contributions

- 8
 - □ The problem
 - CFL-reachability-based alias analysis does not scale in practice.
 - Traditional CFL algorithm has a cubic worst-case complexity
 - CFL-reachability-based alias analysis for C: 8 hours on linux
 - Our contribution & take-away
 - Algorithmic
 - A set of fast CFL-reachability algorithms for alias analysis
 - Practical
 - The first subcubic implementation of CFL-reachability algorithm
 - Less than 80 seconds on the latest linux kernel

Scaling alias analysis for Java

Bidirectness	Туре	Time	Space	Reference
Bidirected	Tree	$O(n\log n\log k)$	$O(n\log n)$	ESOP09
General	Graph	$O(k^{3}n^{3})$	$O(kn^2)$	PODS90, POPL95
General	Graph	$O(kn^3)$	$O(kn^2)$	PLDI04
General	Graph	$O(kn^3/\log n)$	$O(kn^2)$	POPL08
Bidirected	Tree	<i>O</i> (<i>n</i>)	O(n)	Section 3.4
Bidirected	Graph	$O(n + m \log m)$	O(n+m)	Section 3.5
General	Graph	O(mn + Sn)	$O(n^2)$	Section 3.6

[ESOP09] H. Yuan and P. T. Eugster. An efficient algorithm for solving the Dyck-CFL reachability problem on trees

[POPL08] S. Chaudhuri. Subcubic algorithms for recursive state machines

- [PLDI04] J. Kodumal and A. Aiken. The set constraint/CFL reachability connection in practice
- [POPL95] T. W. Reps, S. Horwitz, and S. Sagiv. Precise interprocedural dataflow analysis via graph reachability

[PODS90] M. Yannakakis. Graph-theoretic methods in database theory

Scaling alias analysis for C

PEG Type	Time	Space	Reference
General	$O(n^3)$	$O(n^2)$	POPL08a
General	$O(n^3/\log n)$	$O(n^2)$	POPL08b
General	O(mn + Mn)	$0(n^2)$	Section 5.3
Well-Typed	$O(n(m+\widetilde{M}))$	$O(n^2)$	Section 5.4

[POPL08a] X. Zheng and R. Rugina Demand-driven alias analysis for C[POPL08b] S. Chaudhuri. Subcubic algorithms for recursive state machines



- Scaling alias analysis for Java
 - [Ch.3] Theory: Dyck-CFL-reachability algorithms
 - [Ch.4] Practice: scaling [ECOOP09], [ISSTA11]
- Scaling alias analysis for C
 - [Ch.5] Theory: CFL-reachability algorithms
 - [Ch.6] Practice: scaling [POPL08]
- Conclusion

Chapter 3 Scaling Alias Analysis for Java: Theory

Problem Definition

Dyck-CFL-reachability

- Dyck Grammar
 - $\blacksquare S \to S S \mid a_1 S \bar{a}_1 \mid a_k S \bar{a}_k \mid \varepsilon$
 - Properly matched parentheses: A and \overline{A}
 - Examples: ()([]) and ([)]
- A bidirected graph
 - For each $L(u, v) \in A$, there is a reverse edge $L(v, u) \in \overline{A}$, and vice versa.





Scaling alias analysis for Java

Bidirectness	Туре	Time	Space	Reference
Bidirected	Tree	$O(n\log n\log k)$	$O(n\log n)$	ESOP09
General	Graph	$O(k^{3}n^{3})$	$O(kn^2)$	PODS90, POPL95
General	Graph	$O(kn^3)$	$O(kn^2)$	PLDI04
General	Graph	$O(kn^3/\log n)$	$O(kn^2)$	POPL08
Bidirected	Tree	<i>O</i> (<i>n</i>)	O(n)	Section 3.4
Bidirected	Graph	$O(n + m \log m)$	O(n+m)	Section 3.5
General	Graph	O(mn + Sn)	$O(n^2)$	Section 3.6

[ESOP09] H. Yuan and P. T. Eugster. An efficient algorithm for solving the Dyck-CFL reachability problem on trees

[POPL08] S. Chaudhuri. Subcubic algorithms for recursive state machines

[PLDI04] J. Kodumal and A. Aiken. The set constraint/CFL reachability connection in practice

[POPL95] T. W. Reps, S. Horwitz, and S. Sagiv. Precise interprocedural dataflow analysis via graph reachability

[PODS90] M. Yannakakis. Graph-theoretic methods in database theory

Applications of Dyck-CFL-Reachability

15

Pointer Analysis

- Java: matched field accesses (i.e., read/write)
- C: pointer references and dereferences
- Interprocedural data flow analysis
 - Calls and returns
- Type-based flow analysis
 - In-flow and out-flow

Key Insight

An old result

- [PLDI98] M. Fahndrich, J. S. Foster, Z. Su and A. Aiken. Partial Online
 Cycle Elimination in Inclusion Constraint Graphs
- Basic idea: collapsing nodes in a strongly connected component (SCC)

A new interpretation

- View SCC as a binary relation C
 - i.e., $(u, v) \in C \leftrightarrow u$ and v are in the same SCC
- \square C is reflexive, C is symmetric, and C is transitive
 - *C* is an equivalence class!
- How about Dyck-CFL-Relation?



Dyck-CFL-Relation

17

 \square Consider the binary Dyck-CFL relation D on graph

□ $(u, v) \in D$ iff v is Dyck-reachable from u.

- □ Reflexivity of relation D□ $S \to \varepsilon$
- \Box Symmetry of relation D
- \Box Transitivity of relation D
 - $\square S \to S S$
 - $\blacksquare S \to a_k S \ \overline{a}_k$
- $\square D$ is an equivalence class as the SCC!
 - Basic idea: collapsing nodes in relation D
 E.g., {1, 3, 5}



Tree Algorithm

□ Input:

- A bidirected tree
- Dyck grammar of k kinds of parentheses

Output:

Stratified set representation



Tree Algorithm

19

Major steps:

- Pick a leaf node to start a DFS
- \square Processing opening parentheses a_i
 - Allocate a new node in the lower layer
- \blacksquare Processing closing parentheses \overline{a}_i
 - Collapse/assign the node in the upper layer w.r.t. index i
- The algorithm takes O(n) time and O(n) space



Graph Algorithm

20

An Dyck-path example



6

 a_1

Basic idea

\square Merging node Z_i

Node number Z + index of incoming parentheses i

- Tracking the changes of merging nodes
- Merging the edges of merging nodes

Graph Algorithm

21

An Dyck-path example







- □ A naïve approach
 - Pick a Merging node
 - Search edges
 - **Time complexity** $O(kn^2)$

Graph Algorithm: Node Tracking

□ Fast doubly linked list (FDLL)

- A DLL + a hash table
- \square Insertion, deletion, and query all in expected O(1) time

We use three FDLLs

- A list of merging node candidates
- A list of nodes each merging node merges
- A list of indexes of each node

□ An example

■
$$2_1 \rightarrow 1$$

■ $3_1 \rightarrow 2,4$
■ $4_1 \rightarrow 5$
■ $6_1 \rightarrow 5,7$
 $2_1 \rightarrow 1$
 $a_1 \rightarrow 1$

Graph Algorithm: Edge Merging

\square Pick a merging node Z_i

\square two nodes x and y from the FDLL of z_i

 \square Merging edges from y to x



□ Special degree $\widehat{D}[x]$ of node x

- **a** Always merging from y to x, where $\widehat{D}[y] \leq \widehat{D}[x]$
- **Differences between** $\widehat{D}[x]$ and D[x]

A Running Example

24



Complexity Analysis

Key Claims

- **I** In each edge merging, an edge (x, y) is "moved"
- For the "moved" edge (x, y), either $\widehat{D}[y]$ or $\widehat{D}[x]$ is doubled ■ $\widehat{D}[v] < 2m$ for all $v \in V$
- \blacksquare An edge is "moved" for at most $2\log 2m$ times
- □ The time complexity is $O(n + m \log m)$
- \square The space complexity is O(n+m)



Chapter 4 Scaling Alias Analysis for Java: Practice

Scaling an Alias Analysis for Java

27

- Context-sensitive flow-insensitive and field-sensitive
 - Demand-driven: single-source-single-sink Dyck-CFL-Reachabiltiy
- In our evaluation
 - Traditional Dyck-CFL-reachability algorithm vs. our algorithm
 - All-pairs Dyck-CFL-reachability
 - Context-insensitive flow-insensitive and field-sensitive

[ECOOP09] G. Xu, A. Rountev, M. Sridharan Scaling CFL-Reachability-Based Points-To Analysis Using Context-Sensitive Must-Not-Alias Analysis.

[ISSTA11] D. Yan, G. Xu, A. Rountev Demand-driven context-sensitive alias analysis for Java

Scaling an Alias Analysis for Java

Symbolic points-to graph

- Nodes: abstract heap locations
- Edges: field accesses (read/write)



(a) A code snippet.

(b) Its SPG.

The grammar

■ memAlias → $\overline{f_1}$ memAlias $f_1 | \cdots | \overline{f_k}$ memAlias f_k | memAlias memAlias | ε

Experiments

29

Evaluation benchmarks

DaCapo-2006-10-MR2 and DaCapo-9.12bach

Benchmark	SPG			Tin	ne (s)	Memo	ry (MB)	
	#Nodes	#Edges	#S-pair	#para	CFL	FAST-DYCK	CFL	FAST-DYCK
antlr	16735	13878	19385	1087	37.42	0.041	29.68	20.21
bloat	20320	16224	23080	1197	43.09	0.048	35.09	23.89
chart	44584	36329	50670	2948	253.06	0.119	76.75	52.02
eclipse	17527	14411	20335	1182	42.26	0.042	30.97	21.19
fop	39977	31515	45837	2724	219.53	0.101	67.99	46.08
hsqldb	15015	12693	17615	998	33.39	0.038	27.10	18.22
jython	21615	17381	24487	1240	49.57	0.052	37.20	25.32
luindex	16098	13336	18716	1071	35.15	0.040	28.64	19.45
lusearch	17003	14195	19911	1117	40.22	0.043	30.34	20.73
pmd	18167	14958	20843	1168	40.28	0.046	32.00	21.90
xalan	15030	12645	17608	996	32.93	0.038	26.93	18.21
batik	40273	32052	46225	2565	206.50	0.100	68.77	46.60
eclipse	37531	31889	54471	2221	366.39	0.103	70.82	44.54
jython	63516	49005	85552	2855	947.49	0.163	112.14	72.18
sunflow	39321	31339	45161	2484	196.23	0.096	67.22	45.57
tomcat	45966	37338	63414	3013	622.36	0.124	83.98	53.56

Discussions

30

Understanding the performance

Benchmark	SPG			Tin	ne (s)	Memo	ry (MB)	
	#Nodes	#Edges	#S-pair	#para	CFL	FAST-DYCK	CFL	FAST-DYCK
antlr	16735	13878	19385	1087	37.42	0.041	29.68	20.21
bloat	20320	16224	23080	1197	43.09	0.048	35.09	23.89
chart	44584	36329	50670	2948	253.06	0.119	76.75	52.02
eclipse	17527	14411	20335	1182	42.26	0.042	30.97	21.19
fop	39977	31515	45837	2724	219.53	0.101	67.99	46.08
hsqldb	15015	12693	17615	998	33.39	0.038	27.10	18.22
jython	21615	17381	24487	1240	49.57	0.052	37.20	25.32
luindex	16098	13336	18716	1071	35.15	0.040	28.64	19.45
lusearch	17003	14195	19911	1117	40.22	0.043	30.34	20.73
pmd	18167	14958	20843	1168	40.28	0.046	32.00	21.90
xalan	15030	12645	17608	996	32.93	0.038	26.93	18.21
batik	40273	32052	46225	2565	206.50	0.100	68.77	46.60
eclipse	37531	31889	54471	2221	366.39	0.103	70.82	44.54
jython	63516	49005	85552	2855	947.49	0.163	112.14	72.18
sunflow	39321	31339	45161	2484	196.23	0.096	67.22	45.57
tomcat	45966	37338	63414	3013	622.36	0.124	83.98	53.56

Discussions

31

Interpreting the client analysis

Benchmark	SPG			Tin	ne (s)	Memo	ry (MB)	
	#Nodes	#Edges	#S-pair	#para	CFL	FAST-DYCK	CFL	FAST-DYCK
antlr	16735	13878	19385	1087	37.42	0.041	29.68	20.21
bloat	20320	16224	23080	1197	43.09	0.048	35.09	23.89
chart	44584	36329	50670	2948	253.06	0.119	76.75	52.02
eclipse	17527	14411	20335	1182	42.26	0.042	30.97	21.19
fop	39977	31515	45837	2724	219.53	0.101	67.99	46.08
hsqldb	15015	12693	17615	998	33.39	0.038	27.10	18.22
jython	21615	17381	24487	1240	49.57	0.052	37.20	25.32
luindex	16098	13336	18716	1071	35.15	0.040	28.64	19.45
lusearch	17003	14195	19911	1117	40.22	0.043	30.34	20.73
pmd	18167	14958	20843	1168	40.28	0.046	32.00	21.90
xalan	15030	12645	17608	996	32.93	0.038	26.93	18.21
batik	40273	32052	46225	2565	206.50	0.100	68.77	46.60
eclipse	37531	31889	54471	2221	366.39	0.103	70.82	44.54
jython	63516	49005	85552	2855	947.49	0.163	112.14	72.18
sunflow	39321	31339	45161	2484	196.23	0.096	67.22	45.57
tomcat	45966	37338	63414	3013	622.36	0.124	83.98	53.56

Discussions

32

Demand-driven vs. exhausted analysis

Benchmark	SPG			Time (s)		Memory (MB)		
	#Nodes	#Edges	#S-pair	#para	CFL	FAST-DYCK	CFL	FAST-DYCK
antlr	16735	13878	19385	1087	37.42	0.041	29.68	20.21
bloat	20320	16224	23080	1197	43.09	0.048	35.09	23.89
chart	44584	36329	50670	2948	253.06	0.119	76.75	52.02
eclipse	17527	14411	20335	1182	42.26	0.042	30.97	21.19
fop	39977	31515	45837	2724	219.53	0.101	67.99	46.08
hsqldb	15015	12693	17615	998	33.39	0.038	27.10	18.22
jython	21615	17381	24487	1240	49.57	0.052	37.20	25.32
luindex	16098	13336	18716	1071	35.15	0.040	28.64	19.45
lusearch	17003	14195	19911	1117	40.22	0.043	30.34	20.73
pmd	18167	14958	20843	1168	40.28	0.046	32.00	21.90
xalan	15030	12645	17608	996	32.93	0.038	26.93	18.21
batik	40273	32052	46225	2565	206.50	0.100	68.77	46.60
eclipse	37531	31889	54471	2221	366.39	0.103	70.82	44.54
jython	63516	49005	85552	2855	947.49	0.163	112.14	72.18
sunflow	39321	31339	45161	2484	196.23	0.096	67.22	45.57
tomcat	45966	37338	63414	3013	622.36	0.124	83.98	53.56

Chapter 5 Scaling Alias Analysis for C: Theory

Scaling alias analysis for C

PEG Type	Time	Space	Reference
General	$O(n^3)$	$O(n^2)$	POPL08a
General	$O(n^3/\log n)$	$O(n^2)$	POPL08b
General	O(mn + Mn)	$O(n^2)$	Section 5.3
Well-Typed	$O(n(m+\widetilde{M}))$	$O(n^2)$	Section 5.4

[POPL08a] X. Zheng and R. Rugina Demand-driven alias analysis for C[POPL08b] S. Chaudhuri. Subcubic algorithms for recursive state machines

CFL-Reachability-Based Alias Analysis for C

35

- Precision is equivalent to Andersen-style analysis
 - **[POPL08]** X. Zheng and R. Rugina. **Demand-driven alias analysis for C**.
- The pointer expression graph (PEG)
 A-edges and D-edges



The recursive state machine representation




Traditional Annroach	м	
naamonar Approach	11/1	
	DV	::=
	S_1	::=
The traditional CFL-Reachability	S_2	::=
- CEC normal form	S_1	::=
	S_3	::=
$\blacksquare A \to BC \text{ and } A \to B$	S_3	::=
15 rules	S_3	::=
Adding new summary edge	S_4	::=
	S_3	::=
Y_{lpha}	V	::=
$\tilde{\mathbf{v}}$	V	::=
$(u) - \frac{\lambda}{-} - (v) \xrightarrow{\alpha} (v)$	V	::=
	V	::=
Cubic time worst-case complex	S_1	::=
	•	

DV d

đV

 $S_1 \bar{a}$

 $S_1 M$

 $S_2 \bar{a}$

 $S_2 a$

 $S_1 a$

 $S_3 a$

 $S_3 M$

 $S_4 a$

 S_1

 S_2

 S_3

 S_4

ε

Key Novelty

38

Observed the same-layer-regular property

 $\blacksquare \bar{a}^* a^*$ -reachability





V

d

 \overline{d}

Key Novelty

39

Propagating reachability information only among:

- Original edges in the graph
- Summary edges denoting memory alias (M)



O((m + M)n) worst-case time complexity
 M and m are very sparse in practice!

Alias properties

40

\square *M* and *V* are mutually dependent

$$M ::= \overline{d} V d \tag{1}$$
$$V ::= (M? \overline{a})^* M? (a M?)^* \tag{2}$$

- □ Fact 1: Each M-path is generated by prepending a \overline{d} -edge and appending a d-edge to a V-edge.
- □ Fact 2: Each *M*-path is generated by a path whose $R(p) = \overline{a}^* a^*$, injected with zero or more non-consecutive M-paths.

Alias properties

41

\square *M* and *V* are mutually dependent

$$M ::= \overline{d} V d \tag{1}$$
$$V ::= (M? \overline{a})^* M? (a M?)^* \tag{2}$$

\Box Unique positions of M in V

(a) V-path

$$1 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 2 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 3 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 4 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 5$$
(b) V-path represented by a-edges

$$1 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 2 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 3 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 4 \xrightarrow{\bar{a}} \cdots \xrightarrow{\bar{a}} 5$$

Propagating reachability summaries

\Box Unique positions of M in V

 \Box Unique positions of M in V

- Due to bidirectness of the graph, positions 4 and 5 are symmetric to positions 2 and 1
- Two-phase propagation

Due to bidirectness of the graph, positions 4 and 5 are symmetric to positions 2 and 1

Two-phase propagation



Two-phase propagation

43

\Box Unique positions of M in V



Phase one propagation, starting at positions 3, 4, 5.



Two-phase propagation

44

\Box Unique positions of M in V



Phase two propagation.



Complexity analysis

- 45
 - For each summary edge (u, X, v), our algorithm searches:
 - a-, \overline{a} -, d-, \overline{d} -edges: node degree Δ in the original graph
 - □ *M*-edges: node degree of *M*-edges in the final graph

Total steps

$$\Sigma_{(u,v)}\Delta_v = \Sigma_u(\Sigma_v\Delta_v) = O((m+M)n)$$

Well-typed alias analysis

46

Well-typeness

Type qualifiers of each assignments are compatible
 Example



Basic idea

A bottom-up approach

int x; int *u, *w, *y, *z; int **v; u = *v; *v = w; v = w; v = &y; y = &x; z = y;

(a) A code snippet.



Computing bottom-layer reachability

- 48
 - □ With no *D*-edges, with only *A*-edges
 - Same-layer-regular property
 - A regular language reachability \bar{a}^*a^*
 - Formulated as a dynamic path problem [SODA90]
 - A reachability matrix $m_{u_r v_q}$
 - A spanning tree $T(u_r)$ associated with each node
 - Amortized O(n) time for each edge insertion

[SODA90] A. L. Buchsbaum, P. C. Kanellakis, and J. S. Vitter. A data structure for arc insertion and regular path finding.

Computing bottom-layer reachability

49

Finite state automata representation



Processing *a*-edges and \overline{a} -edges

Edge inserted	$m_{u_rv_q}$ updated	$T(x_t)$ affected
	$m_{u_1v_1}$	$T(x_0)$
(u, a, v)	$m_{u_0v_1}$	$T(x_0)$
	$m_{u_1v_1}$	$T(x_1)$
(u, \bar{a}, v)	$m_{u_0v_0}$	$T(x_0)$

Computing bottom-layer reachability

50

Updating reachability information



An example

51

Before edge insertion



(a) The input PEG.

(b) The reachability spanning trees.

An example

52

 \square Inserting (1, a, 4) \Box Updating $m_{1_04_1}$ and $m_{1_14_1}$ \square For each x_t , the pruned cc

 $T(2_0)$ $T(1_0)$ $T(2_1)$ $T(3_0)$ 3_0 1_0 2_0 (**3**1) $(\hat{\mathbf{l}}_{\mathbf{l}})$ $[2_0]$ 2_0 4_0 1_1 $(3_1$ 1_1 3_1

 $(5_1$







(a) New reachability trees.

Computing reachability among layers

- 53
 - Layers connected via *d*-edges and *d*-edges
 Assume layer *l* 1 is computed, consider layer *l*
 - □ For each (x, d, w), search v at layer l 1
 - Update $m_{u_q w_r}$ using $m_{v_q x_r}$



Complexity analysis (sketch)

- 54
 - \Box *a*-, \overline{a} -, *d* and \overline{d} -edges are original edges
 - \square For each a-, \overline{a} edges on layer l
 - **Regular language reachability** (\bar{a}^*a^*)
 - **Each** edge insertion in amortized O(n) time [SODA90]
 - □ For each d-, \overline{d} edges (e.g., (x, d, w))
 - Update $m_{u_q w_r}$ using $m_{v_q x_r}$
 - **Takes** O(n) time
 - □ The whole algorithm takes $O(n(m + \tilde{M}))$ time

[SODA90] A. L. Buchsbaum, P. C. Kanellakis, and J. S. Vitter. A data structure for arc insertion and regular path finding.

Chapter 6 Scaling Alias Analysis for C: Practice

Scaling an Alias Analysis for C

- 56
 - Context-insensitive flow-insensitive and field-insensitive
 - Demand-driven: single-source-single-sink CFL-Reachability
 - In our evaluation
 - Traditional CFL-reachability algorithms vs. our algorithm
 - All-pairs CFL-reachability
 - Context-insensitive flow-insensitive and field-insensitive

[POPL08] X. Zheng and R. Rugina Demand-driven alias analysis for C

Implementation

- □ Front end: gcc-4.6.3
- Applying the Four-Russains' trick:
 - Known technique in subcubic CFL-reachability algorithm, assuming RAM model with word size of $\theta(\log n)$
 - Store n elements using fast set in $O([n \log n])$ words
 - **Set difference** Diff(X, Y) in $O(\lceil n \log n \rceil + v)$ time
 - Directly applicable to our algorithm
- Connect component decomposition

Applying the Four-Russians' Trick

58

Propagating reachability summary



(a). Phase one. (b) Phase two.

Modifications:

foreach $w \in OUT(v, \alpha)$ do if $(u, Y_{\alpha}, w) \notin G$ then

becomes

foreach $w \in diff(OUT(v, \alpha), OUT(u, Y_{\alpha}))$ do

Results

59

Benchmark programs

Program	SLOC	#	PEG		
		#procedures	#nodes	#edges	
Gdb-7.5.1	1,828K	10,536	649,564	1,219,788	
Emacs-24.2	254K	3,626	687,691	1,290,200	
Insight-6.8-1 a	1,742K	10,507	787,289	1,494,168	
Gimp-2.8.4	702K	17,842	872,681	1,675,546	
Ghostscript-9.07	851K	12,211	1,198,753	2,368,086	
Wine-1.5.25	2,306K	70,923	4,652,983	8,472,950	
Linux-3.8.2	10,601K	138,095	12,807,645	23,398,670	

Connected component decomposition

60

Connected component information

Program	#CC -	PEG in	Proc.	PEG in CC		
		Max.	Avg.	Max.	Avg.	
Gdb-7.5.1	66,154	5,162	61.65	4,350	9.82	
Emacs-24.2	67,608	22,654	189.66	15,690	10.17	
Insight-6.8–1 a	75,373	6,273	74.93	4,350	10.45	
Gimp-2.8.4	79,572	3,599	48.91	2,693	10.97	
Ghostscript-9.07	87,768	8,573	98.17	7,025	13.66	
Wine-1.5.25	537,370	20,008	65.61	5,106	8.66	
Linux-3.8.2	1,449,718	7,305	92.75	4,755	8.83	

Results

Performance comparisons

Pro errore	Time (seconds)			Memory (MBs)		
rogram	Cubic	Subcubic	Our	Cubic	Subcubic	Our
Gdb-7.5.1	2405.62	10.83	5.69	150.54	102.6	56.85
Emacs-24.2	54781.9	128.03	90.8	1095.5	1908.16	659.16
Insight-6.8-1 a	665.06	8.65	3.31	116.29	149.57	56.31
Gimp-2.8.4	662.27	8.19	3.01	56.84	51.04	22.58
Ghostscript-9.07	4209.84	28.67	16.23	256.4	271.51	140.79
Wine-1.5.25	8234.29	64.07	21.19	451.28	1448.21	76.14
Linux-3.8.2	31997	160.81	73.45	236.44	197.98	66.09

Results

62

□ Final graph density

Program	#orig. edges	#IZ adaga		#Final edges
		#v-edges	#M-edges	CFL-reachability
Gdb-7.5.1	1,219,788	12,904,372	356,075	29,961,321
Emacs-24.2	1,290,200	49,011,799	758,925	112,772,537
Insight-6.8–1 a	1,494,168	12,560,471	432,657	27,573,822
Gimp-2.8.4	1,675,546	16,809,343	518,511	33,151,263
Ghostscript-9.07	2,368,086	35,910,829	705,121	76,022,402
Wine-1.5.25	8,472,950	79,613,731	2,769,135	161,447,212
Linux-3.8.2	23,398,670	234,930,383	6,272,658	482,622,025



Summary

- Fast Dyck-CFL-reachability algorithm to scale alias analysis for Java
- Fast CFL-reachability algorithm to scale alias analysis for C
- Built practical tools to handle real-world programs
 - Soot for Java
 - Gcc for C

Thanks!



Alias on SPG

- $\Box x \leftarrow z \text{ and } w \rightarrow y$
 - x=z.f and y=w.f
 - x=z.f and w.f=y
 - z.f=x and y=w.f
 - z.f =x and w.f=y

Traditional CFL-Reachability algorithm

Algorithm 1: CFL-Reachability Algorithm. **Input** : Edge-labeled directed graph G = (V, E); normalized $CFG = (\Sigma, N, P, S);$ Output: the set of summary edges; 1 add E to W: 2 foreach production $A \to \varepsilon \in P$ do for each *node* $v \in V$ do 3 if $(v, A, v) \notin G$ then 4 insert (v, A, v) to G and to W; 5 6 while $W \neq \emptyset$ do $(i, B, j) \leftarrow \text{SELECT-FROM}(W);$ $\mathbf{7}$ for each production $A \rightarrow B \in P$ do 8 if $(i, A, j) \notin G$ then 9 insert (i, A, j) to G and to W; 10for each production $A \rightarrow BC \in P$ do 11 for each $k \in OUT(j, C)$ do 12if $(i, A, k) \notin G$ then 13insert (i, A, k) to G and to W; $\mathbf{14}$ for each production $A \to CB \in P$ do 15for each $k \in IN(i, C)$ do 16if $(k, A, j) \notin G$ then 17insert (k, A, j) to G and to W; 18

Dyck-CFL-Reachability tree algorithm

69

Procedure 3: Add(v, e) to add a node v to STRATIFIED-SETS according to the directed edge e = (u, v).

```
1 if \mathcal{L}(u, v) \in A then
     let a_i = \mathcal{L}(u, v)
 \mathbf{2}
     Set[v] = curset
 3
     Up[curset][a_i] = Find(u)
 4
       curset ++
 5
 6 if \mathcal{L}(u,v) \in \overline{A} then
       let \bar{a}_i = \mathcal{L}(u, v)
 7
       if Up[Find(u)][a_i] does not exist then
 8
           Set[v] = curset
 Q
           Up[Find(u)][a_i] = curset
10
           curset ++
11
       else
12
         Set[v] = Up[Find(u)][a_i]
13
```

Well-typed algorithm

Procedure 13: Add (u, \mathcal{L}, v) to insert an edge (u, \mathcal{L}, v) .

1 for each $x \in V[l(v)]$ do if $\mathcal{L}(u, v) == a$ then $\mathbf{2}$ if $m_{x_0u_1} == 1$ and $m_{x_0v_1} \neq 1$ then 3 Mix (x_0, v_1, u_1, v_1) 4 if $m_{x_0u_0} == 1$ and $m_{x_0v_1} \neq 1$ then 5 Mix (x_0, v_1, u_0, v_1) 6 if $m_{x_1u_1} == 1$ and $m_{x_1v_1} \neq 1$ then 7 Mix (x_1, v_1, u_1, v_1) 8 if $\mathcal{L}(u, v) == \bar{a}$ then 9 if $m_{x_0u_0} == 1$ and $m_{x_0v_0} \neq 1$ then 10Mix (x_0, v_0, u_0, v_0) 11 if $\mathcal{L}(u, v) == d$ then 12for each $q \in \{0, 1\}, r \in \{0, 1\}, s \in \{0, 1\}$ do 13 if $m_{v_{a}x_{r}} == 1$ and u(x) exists then 14 $w \leftarrow u(x)$ 15if $m_{u_s w_s} \neq 1$ then $m_{u_s w_s} \leftarrow 1$ and insert w_s as a child of $T(u_s)$ 16

Well-typed algorithm

71

Procedure 14: $Mix(x_t, v_q, i_l, j_m)$ to merge trees.

1 insert j_m in $T(x_t)$ as a child of i_l 2 $m_{x_t j_m} \leftarrow 1$ 3 foreach child k_n of $j_m \in T(v_q)$ do 4 $\begin{bmatrix} \text{if } m_{x_t k_n} \neq 1 \text{ then} \\ & & \\ \text{5} \end{bmatrix} \begin{bmatrix} \text{Mix } (x_t, v_q, j_m, k_n) \end{bmatrix}$

Algorithm 15: Pointer analysis algorithm for well-typed C.

Input : Edge-labeled bidirected PEG G = (V, E); **Output**: the reachability matrix M

1 run pre-process pass described in Section 5.4.1

2 Init ()

3 for $k \leftarrow bottom$ to top do

- 4 for each $(i, a, j), (i, \overline{a}, j) \in E_k$ do Add $(i, \mathcal{L}(i, j), j)$
- 5 for each $(i, d, j) \in E_k$ do Add $(i, \mathcal{L}(i, j), j)$

General Dyck-CFL-Reachability

72

Procedure 7: Add(i, j) to insert an edge (i, j).

1 if $m_{ij} == 0$ then // there are no previous path from i to j2 for $x \leftarrow 1$ to n do 3 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 4 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 5 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 4 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 5 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 6 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 7 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 8 lif $m_{xi} \neq 0$ and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_{xj} == 0$ then 9 lif m_{xi} \neq 0 and $m_$

Procedure 8: Meld(x, j, u, v) to merge trees.

// update by means of
Algorithm 9: Dyck-CFL-Reachability Algorithm. **Input** : Edge-labeled directed graph G = (V, E); Output: the set of summary edges; **1** initialize W to be empty **2** foreach $(i, a_i, j) \in E$ do insert (i, A_i, j) to G and to W **3** for each $(i, \bar{a}_i, j) \in E$ do insert (i, A_i, j) to G and to W 4 while $W \neq \emptyset$ do $(i, B, j) \leftarrow \text{SELECT-FROM}(W)$ 5 if $B == \overline{A_i}$ then 6 for each $k \in IN(i, a_i)$ do 7 if $(k, S, j) \notin G$ then 8 Add (k, j)9 insert (k, S, j) to G and to W 10if $B == A_i$ then 11 for each $k \in OUT(j, \bar{a}_i)$ do 12 if $(i, S, k) \notin G$ then 13 Add (k, j) $\mathbf{14}$ insert (i, S, k) to G and to W 15if B == S then 16 for each $k \in IN(i, a_i)$ do 17if $(k, A_i, j) \notin G$ then $\mathbf{18}$ insert (k, A_i, j) to G and to W 19for each $k \in OUT(j, \bar{a}_i)$ do $\mathbf{20}$ if $(i, \overline{A}_i, k) \notin G$ then $\mathbf{21}$ insert (i, \overline{A}_i, k) to G and to W $\mathbf{22}$

General Dyck-CFL-Reachability

 \square Amortized time *a*

$$a = t + \sum_{v \in V} \varphi'(v) - \sum_{v \in V} \varphi(v).$$

 \Box Total running time for *m* insertion

$$\sum_{i=1}^m t_i = \sum_{i=1}^m a_i + \sum_{v \in V} \varphi(v) - \sum_{v \in V} \varphi'(v).$$

 Define vis(x) = {(u, v) ∈ E|u is a descendant of x}
 Define potential function φ(x) = −(|vix(x)| + |T(x)|)

General Dyck-CFL-Reachability

75

- □ Meld() searches h_1 edges in T(j) and adds h_2 edges to T(x)
- A net decrease of potential
 - Updating T(x) does not sear
 - Each searched edge is beco
 - **T**(x) is increased by h_2
- □ Inserting each edge to T(x)
- As a result for each x, each amortized time
 - We have n such x nodes

[TCS86] G.F. Italiano Amortized efficiency of a path retrieval data structure

