

Shape Analysis

Rupesh Nasre.

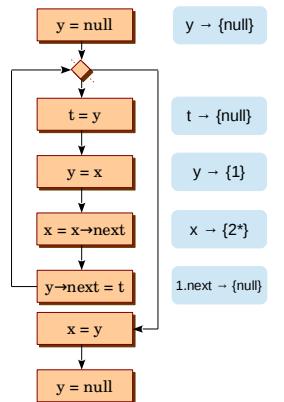
CS6843 Program Analysis
IIT Madras
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Limitations of Pointer Analysis

```
listReverse(List x) {
    assert("x is an acyclic singly linked list");

    for (y = null; x;) {
        t = y;
        y = x;
        x = x->next;
        y->next = t;
    }
    x = y;
    t = null;
    y = null;
}
```

We model the list as a two node structure.



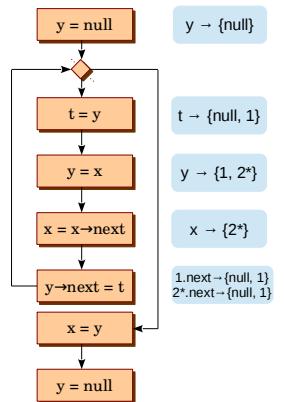
Outline

- Limitations of pointer analysis
- Identify lists
- Identify trees, DAGs, cyclic graphs
- Identifying rotations
- List reversal and other transformations

Limitations of Pointer Analysis

```
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    assert("x is an acyclic singly linked list");

    for (y = null; x;) {
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        x = x->next;
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    }
    x = y;
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}
```



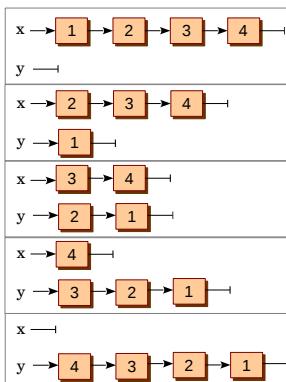
Limitations of Pointer Analysis

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    for (y = null; x;) {
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We want to check if x points to a singly linked list at the end of listReverse.
That is,
x -> {4}, 4.next -> {3}, ..., 1.next -> {null}

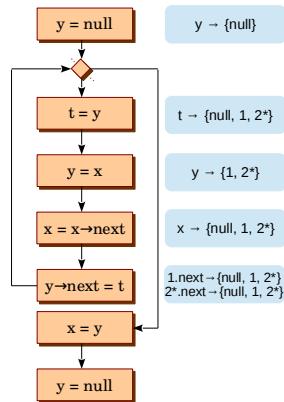
```



Limitations of Pointer Analysis

```
listReverse(List x) {
    assert("x is an acyclic singly linked list");

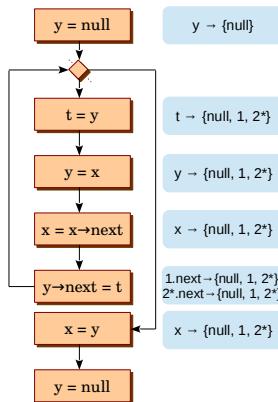
    for (y = null; x;) {
        t = y;
        y = x;
        x = x->next;
        y->next = t;
    }
    x = y;
    t = null;
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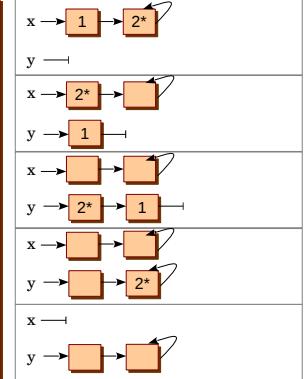


Shape Analysis

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```

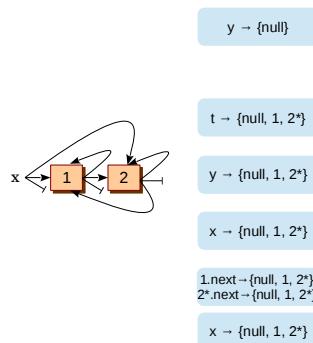
Maintain additional information with 2^* that it is acyclic.
Use the fact that node removal maintains acyclicity.



Limitations of Pointer Analysis

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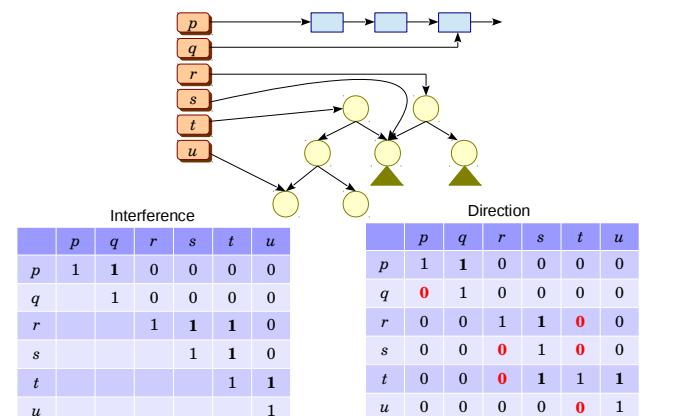
Tree, DAG, Cycle?

- Maintains three data structures:
 - Interference matrix: encodes common reachability
 - Direction matrix: encodes direct reachability
 - Shape
- Performs iterative data-flow analysis to update D, I and shape information

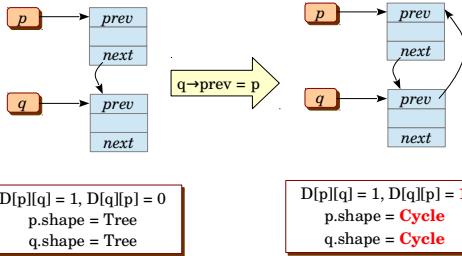
Shape Analysis

- Identify structural / topological properties of a data structure under manipulation.
- Usually categorized as slist, tree, DAG or cycle.
- Precision reduces along slist \rightarrow tree \rightarrow DAG \rightarrow cycle.

Tree, DAG, Cycle?



Shape Estimation



Inference Rules

p = malloc(...)	D_kill and I_kill sets same as for allocation statement.
p = q	D_gen = {D[s][p] D[s][q] and s ≠ p} U {D[p][s] D[q][s] and s ≠ p} U {D[p][p] D[q][q]}
p = q->f	I_gen = {I[p][s] I[q][s] and s ≠ p} U {I[p][p] I[q][q]}
p = &(q->f)	
p = q op k	
p = null	
p->f = q	
p->f = null	
	p.shape = q.shape

The implementation should create new D/I matrices from their current copies. In-situ update would lead to unsound or imprecise analysis.

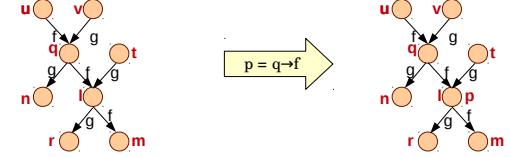
Inference Rules

p = malloc(...)
p = q
p = q->f
p = &(q->f)
p = q op k
p = null

p->f = q
p->f = null

Inference Rules

p = malloc(...)	D_kill and I_kill sets same as for allocation statement.
p = q	D_gen = {D[s][p] I[s][q] and s ≠ p} U {D[p][s] D[q][s] and s ≠ p and s ≠ q} U {D[p][q] q.shape == Cycle} U {D[p][p] D[q][q]}
p = q->f	I_gen = {I[p][s] I[q][s] and s ≠ p} U {I[p][p] I[q][q]}
p = &(q->f)	
p = q op k	
p = null	
p->f = q	
p->f = null	
	p.shape = q.shape



Inference Rules

p = malloc(...) D_kill = {D[p][s] | D[p][s] == 1} U {D[s][p] | D[s][p] == 1} I_kill = {I[p][s] | I[p][s] == 1}
p = q
p = q->f
p = &(q->f)
p = q op k
p = null

p->f = q
p->f = null

Inference Rules

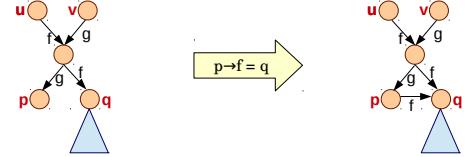
p = malloc(...)	Processing is the same as for p = q statement. This means the analysis loses field-sensitivity.
p = q p = q->f p = &(q->f) p = q op k p = null p->f = q p->f = null	A former work from IITK (Dasgupta, Karkare, Reddy) addresses this issue.

Inference Rules

$p = \text{malloc}(\dots)$	$D_{\text{kill}} \text{ and } I_{\text{kill}}$ sets same as for allocation statement.
$p = q$	$D_{\text{gen}} = \{\}$
$p = q \rightarrow f$	$I_{\text{gen}} = \{\}$
$p = \&(q \rightarrow f)$	$p.\text{shape} = \text{Tree}$
$p = q \text{ op } k$	
$\mathbf{p = null}$	
$p \rightarrow f = q$	
$p \rightarrow f = \text{null}$	

Inference Rules

$p = \text{malloc}(\dots)$	$D_{\text{kill}} = \{\}, I_{\text{kill}} = \{\}$
$p = q$	$D_{\text{gen}} = \{D[r][s] \mid D[r][p] \text{ and } D[q][s]\}$
$p = q \rightarrow f$	$I_{\text{gen}} = \{I[r][s] \mid D[r][p] \text{ and } I[q][s]\}$
$p = \&(q \rightarrow f)$	$D[q][p] \text{ and } D[s][q] \Rightarrow s.\text{shape} = \text{Cycle}$
$p = q \text{ op } k$	$D[q][p] \text{ and } D[s][p] \Rightarrow s.\text{shape} = \text{Cycle}$
$p = \text{null}$	$!D[q][p] \text{ and } D[s][p] \text{ and } I[s][q] \text{ and } q.\text{shape} == \text{Tree}$ $\Rightarrow s.\text{shape} = \max(s.\text{shape}, \text{DAG})$
$\mathbf{p \rightarrow f = q}$	$!D[q][p] \text{ and } D[s][p] \text{ and } q.\text{shape} != \text{Tree}$ $\Rightarrow s.\text{shape} = \max(s.\text{shape}, q.\text{shape})$
$p \rightarrow f = \text{null}$	



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$p \rightarrow f = \text{null}$	

Can you improve precision?

Inference Rules

$p = \text{malloc}(\dots)$	$D_{\text{kill}} = \{\}, I_{\text{kill}} = \{\}$
$p = q$	$D_{\text{gen}} = \{D[r][s] \mid D[r][p] \text{ and } D[q][s]\}$
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$p = \&(q \rightarrow f)$	$D[q][p] \text{ and } D[s][q] \Rightarrow s.\text{shape} = \text{Cycle}$
$p = q \text{ op } k$	$D[q][p] \text{ and } D[s][p] \Rightarrow s.\text{shape} = \text{Cycle}$
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$p \rightarrow f = \text{null}$	

Tree	DAG	Cycle
Tree	DAG	Cycle
DAG	DAG	Cycle
Cycle	Cycle	Cycle

$\max(\text{shape1}, \text{shape2})$

Inference Rules

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For $\max()$ consider $D[v][r] == 1$.

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$p = q \rightarrow f$	$I_{\text{gen}} = \{\}$
$p = \&(q \rightarrow f)$	No changes to the shape of p .

Example

```
listReverse(List x) {
    assert("x is an acyclic singly linked list");

    for (y = null; x;) {
        t = y;
        y = x;
        x = x->next;
        y->next = t;
    }
    x = y;
    t = null;
    y = null;
}
```

Interference		
x	y	t
x	0	0
y		0
t		

Direction		
x	y	t
x	1	0
y	0	0
t	0	0

Shape		
x	y	t
x	tree	
y	tree	
t	tree	

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Direction		
x	y	t
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y	1	1
t	0	0

Shape		
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x	tree	
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x	y	t
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t		

Direction		
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y	0	0
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Shape		
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Shape		
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tree		
tree		
tree		

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Direction		
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Shape		
x	y	t
tree		
cycle		
cycle		

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Shape		
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Classwork

```
p = malloc(10);
p->f1 = null;
q = p->f2;
q = &(r->f2);
q->f2 = p;
```

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Shape		
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x	tree	
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Improvements

- Field-sensitivity
- Heap modeling
- Path-sensitivity

Example

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Shape		
x	y	t
x	cycle	
y	cycle	
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Summary

- Shape analysis helps several transforms.
- Existing techniques often trade off precision for efficiency.
- We are still far away from a precise and scalable analysis.

