

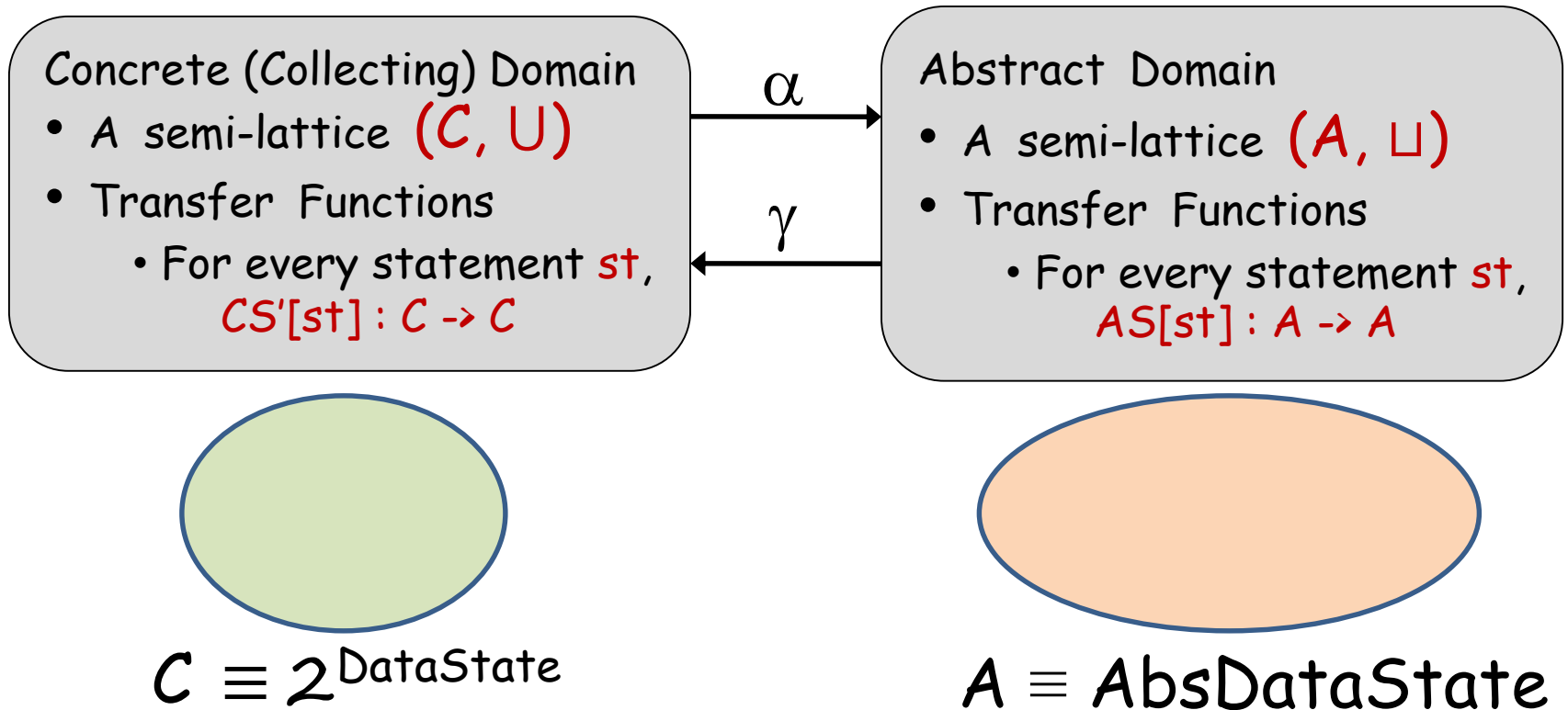
# Pointer Analysis

## Lecture 2

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*Correctness and precision of  
Algorithm A*

# Enter: The French Recipe (Abstract Interpretation)



# Points-To Analysis (Abstract Interpretation)

$$\alpha(Y) = \lambda p. \{x \mid \text{exists } s \text{ in } Y. s(p) == x\}$$

$$\gamma(a) = \{(p \rightarrow v, q \rightarrow w) \mid v \in a(p), w \in a(q)\}$$

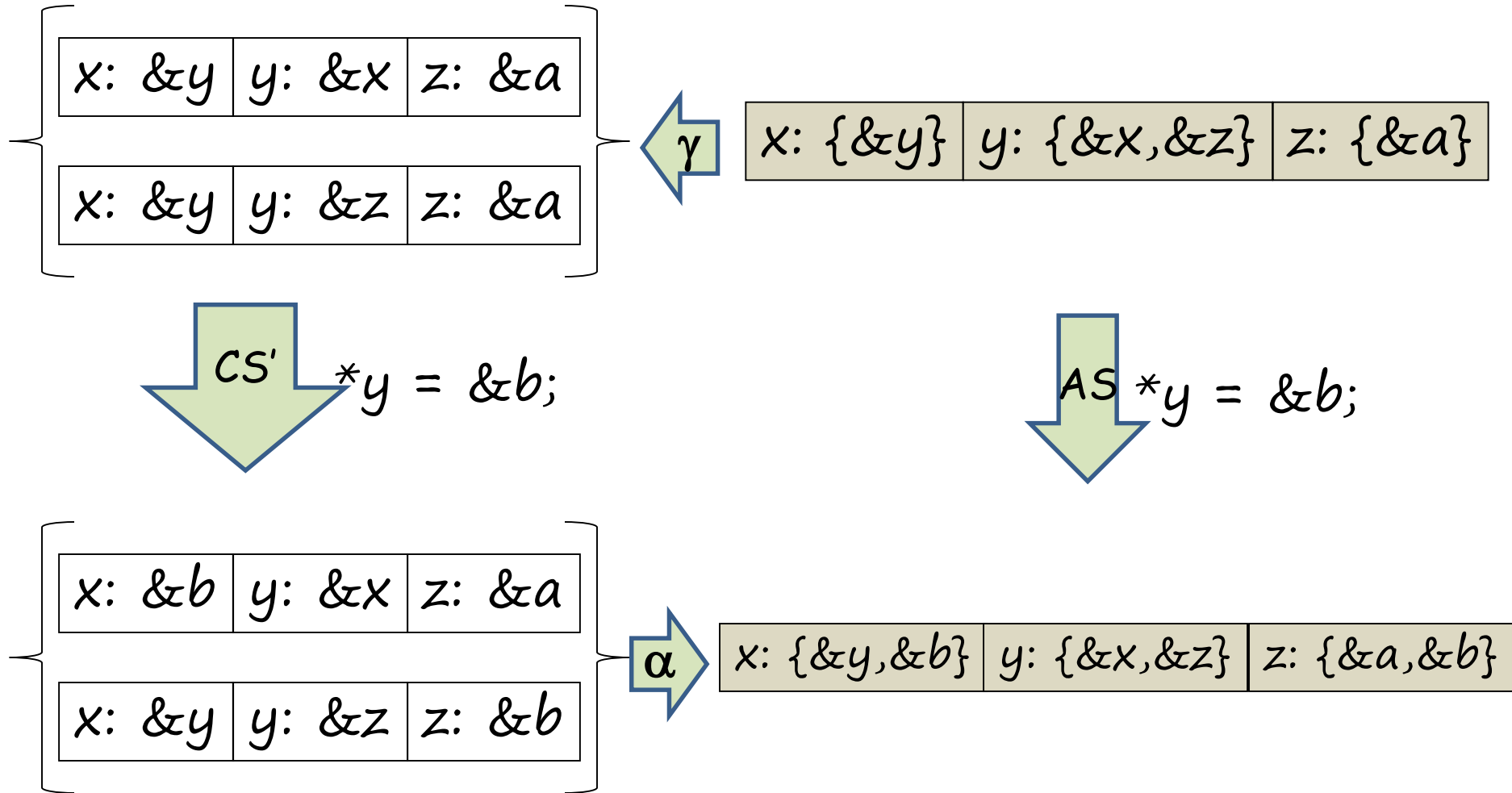
(assuming  $p, q$  are the pointer variables in the program)

# Approximating Transformers: Correctness Criterion

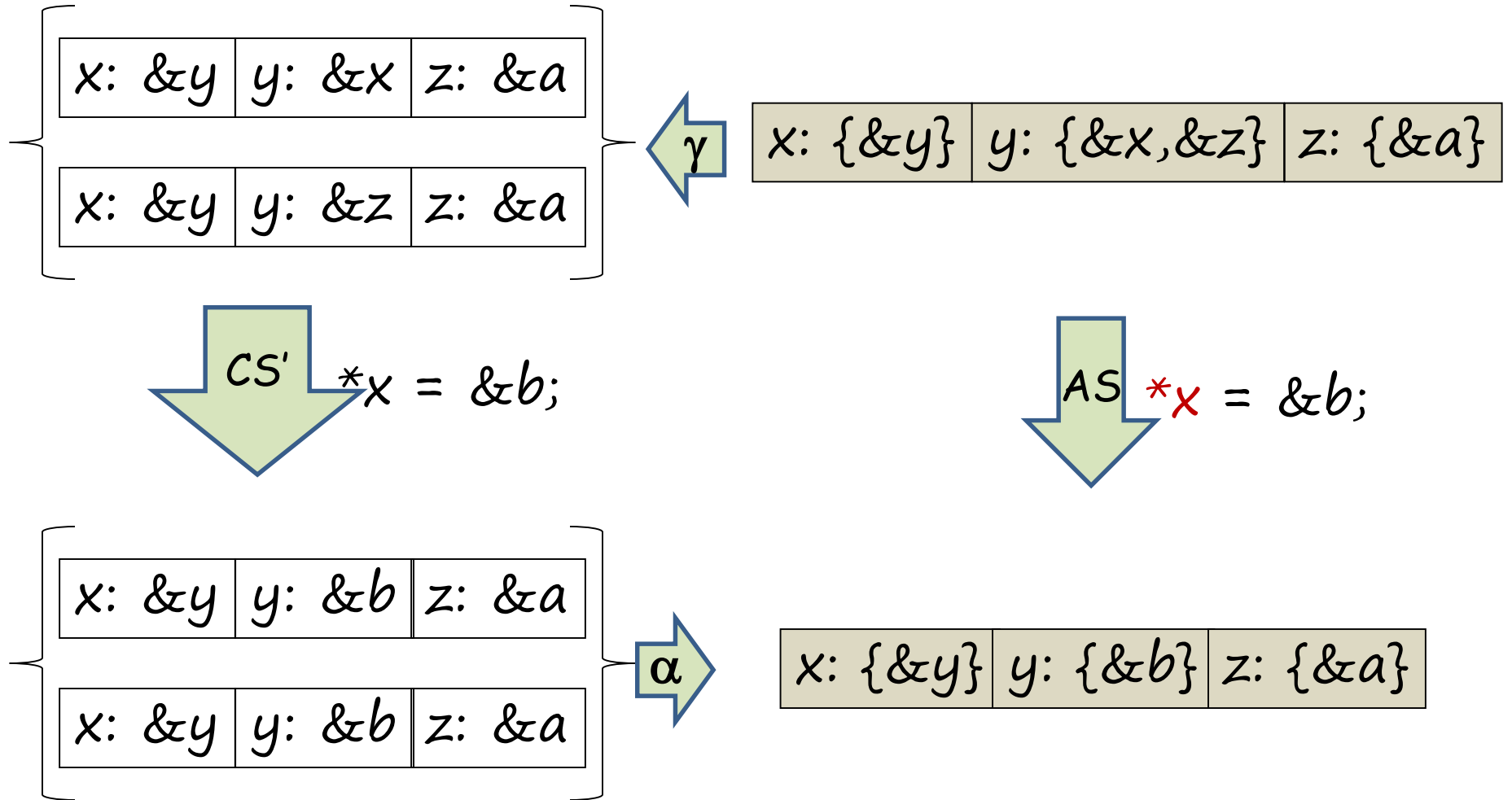
It can be shown that for any  
statement  $st$ , and for any  $a1 \in A$

$$AS[st](a1) \geq \alpha (CS[st](\gamma(a1)))$$

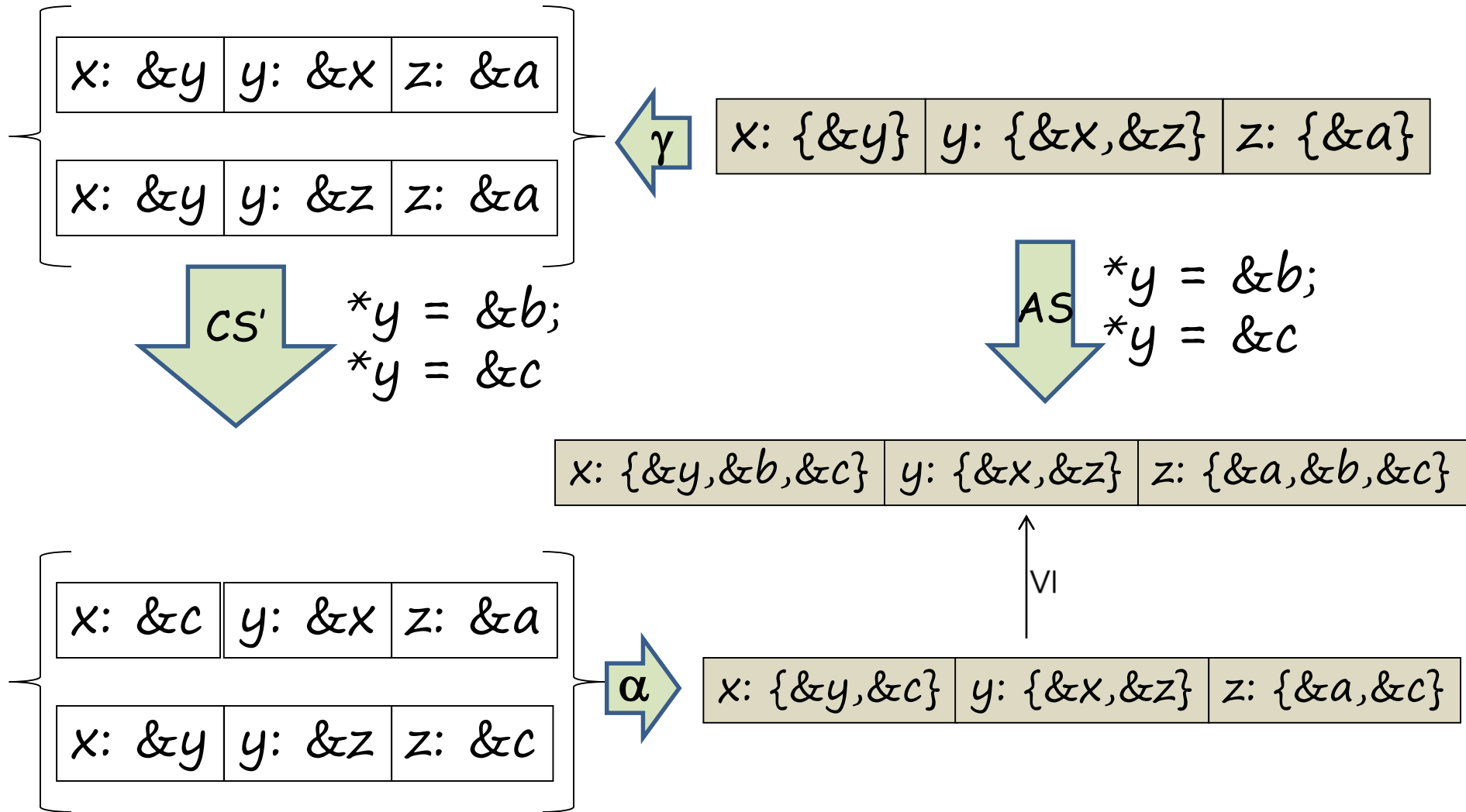
# Transfer function illustration



# Transfer function illustration



# Transfer function illustration





# Is The Precise Solution Computable?

- Claim: The set  $RS(u)$  of reachable concrete states (for our language) is precisely computable.
  - (However, Algorithm A is imprecise)
- Note: This is true for any collecting semantics with a finite state space.
- This is true only for restricted language!

# Precise Points-To Analysis: Computational Complexity

- What's the complexity of the least-fixed point computation using the collecting semantics?
- The worst-case complexity of computing reachable states is exponential in the number of variables.
  - Can we do better?
- Theorem: Computing precise may-point-to is PSPACE-hard even if we have only two-level pointers.

# Precise Points-To Analysis: Caveats

- Theorem: *Precise may-alias analysis is undecidable in the presence of dynamic memory allocation.*
  - Add “*x = new/malloc ()*” to language
  - State-space becomes infinite
- Digression: *Integer variables + conditional-branching* involving integer variables also makes any precise analysis undecidable.

# Dynamic Memory Allocation

- $s: x = \text{new} () / \text{malloc} ()$
- Assume, for now, that allocated object stores one pointer
  - $s: x = \text{malloc} ( \text{sizeof}(\text{void}^*) )$
- Introduce a pseudo-variable  $V_s$  to represent objects allocated at statement  $s$ , and use previous algorithm
  - treat  $s$  as if it were “ $x = \&V_s$ ”
  - also track possible values of  $V_s$
  - allocation-site based approach

# $\alpha$ in the presence of pseudo variables

$\alpha(Y) = \setminus p. \{x \mid \text{exist } s \text{ in } Y \text{ such that } s(y) = z$   
such that:

( $p$  is a normal variable and  $y$  is  $p$ ) OR

( $p$  is  $V_r$  and  $y$  is an address  
allocated at site  $r$  )

AND

( $x$  is a normal variable and  $x$  is  $z$ ) OR

( $x$  is  $V_t$  and  $z$  is an address  
allocated at site  $t$  ) }

(For simplicity, assume that the set of all concrete addresses is partitioned upfront among all allocation sites in the program)

# $\gamma$ in the presence of pseudo variables

$\gamma(a) = \{s \mid \forall \text{ normal variables } p:$

$s(p) = x \wedge x \text{ is a normal variable} \wedge x \in a(p), \text{ OR}$

$s(p) = y \wedge y \text{ is an address allocated at}$   
 $\text{site } t \wedge V_t \in a(p), \text{ AND}$

$\forall \text{ pseudo-variables } V_r: \forall \text{ addresses } y \text{ allocated at } V_r:$

$s(y) = x \wedge x \text{ is a normal variable} \wedge x \in a(V_r), \text{ OR}$

$s(y) = z \wedge z \text{ is an address allocated at}$   
 $\text{site } t \wedge V_t \in a(V_r)\}$

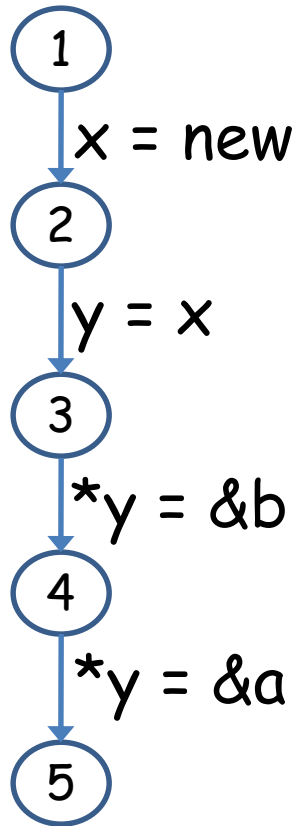
# Dynamic Memory Allocation: Example

```
x = new;
```

```
y = x;
```

```
*y = &b;
```

```
*y = &a;
```



# Example illustrating need for weak updates on summary objects

- Key aspect:  $V_s$  represents a set of objects (locations), not a single object
  - referred to as a **summary** object (node)
  - if  $x \rightarrow \{V_s\}$ , to be safe “\*x = ..” still needs weak update!
- If example fragment in previous slide is placed in a loop, then solution at vertex 4 would be wrong unless weak-update is adopted.



# Handling structures/objects

- Concrete state maps each variable to a variable or address of an object or null, and for each object  $o$  and each field  $f$  of  $o$ , maps  $(o, f)$  to the address of some other object (or to null)
- Let  $PseudoVar$  be the set of pseudo-variables, let  $Var' = Var \cup PseudoVar \cup \{null\}$
- Each  $AbsDataState$  maps each variable to a subset of  $Var'$ , and also maps  $(V_s, f)$  to a subset of  $Var'$  for each  $V_s$  and for each field  $f$  in struct allocated at site  $s$

# Handling structures/objects

- Transfer function of an update statement " $v \rightarrow f = w$ " should perform weak update if  $v$  points to a singleton and that element is a pseudo-variable.
  - That is, if  $v \rightarrow \{V_s\}$  in incoming abstract state, then add points-to set of  $w$  to points-to set of  $(V_s, f)$

# Inter-procedural analysis

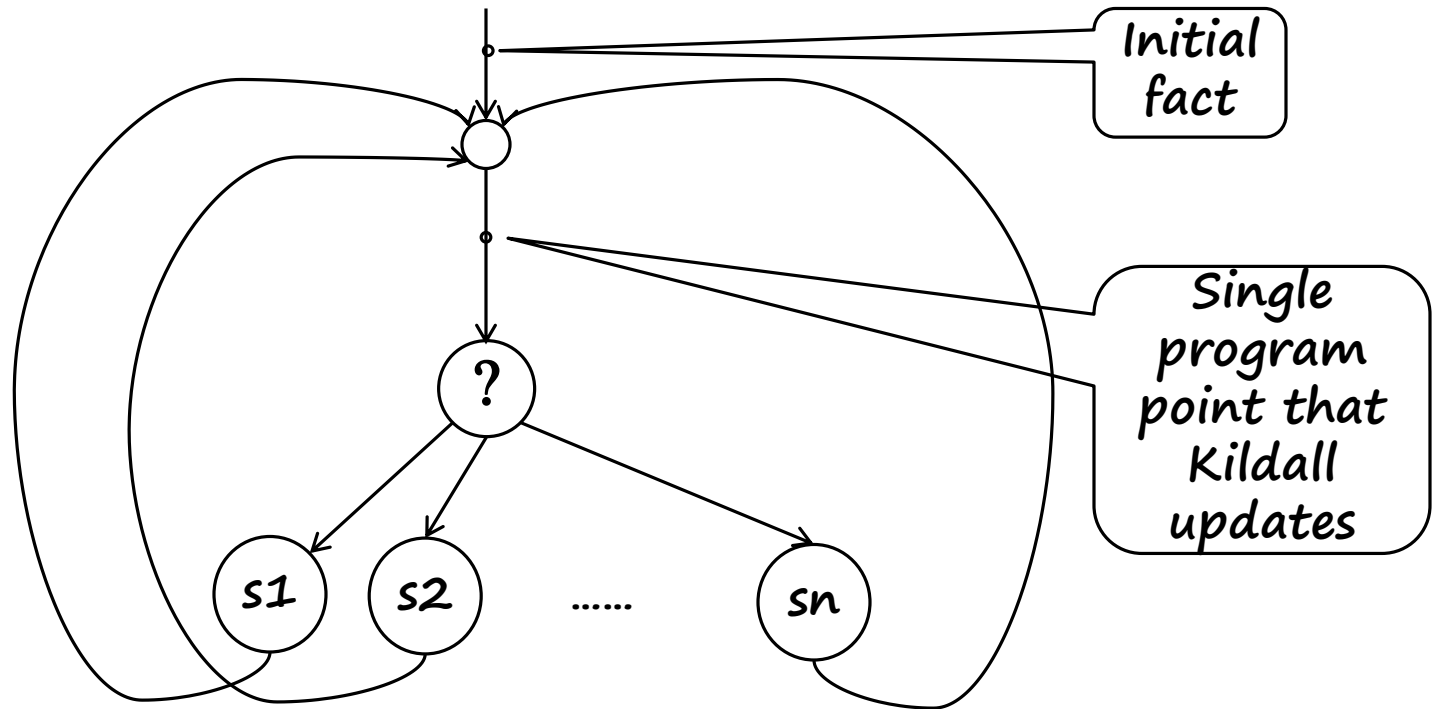
- Context-sensitivity can be achieved using standard techniques
- Indirect (virtual) function call sites need to be resolved to candidate functions using points-to analysis. And points-to analysis needs calls to be resolved! Therefore, the two have to happen hand in hand.

# Andersen's Analysis

- A *flow-insensitive* analysis
  - computes a single points-to solution, which over-approximates points-to solutions at all program points
  - ignores control-flow – treats program as a set of statements
  - equivalent to collapsing the given program to have a single program point, and then applying Algorithm A on it.

# Andersen's Analysis

If program has statements  $s_1, s_2, \dots, s_n$ , then create collapsed CFG as follows:



After algorithm terminates, final solution at the single program point over-approximates result computed by flow-sensitive analysis at any point

# Example: Andersen's Analysis

```
x = &a;
```

```
*x = &w;
```

```
y = x;
```

```
x = &b;
```

```
*x = &t;
```

```
z = x;
```

# Notes about Andersen's Analysis

- Strong updates never happen in Andersen's analysis!
  - If  $x \rightarrow \{y\}$  and  $y \rightarrow \{w\}$  before we process statement “ $*x = \&z$ ”, then even if transfer function returns  $y \rightarrow \{z\}$ , due to subsequent join,  $y$  will point to  $\{w, z\}$  after this step.
- Flow-insensitive style can be adopted for **any** analysis, not just for pointer analysis

# Why Flow-Insensitive Analysis?

- Reduced space requirements
  - a single points-to solution
- Reduced time complexity
  - no copying of points-to facts
    - individual updates more efficient
  - a cubic-time algorithm
- Scales to millions of lines of code
  - most popular points-to analysis

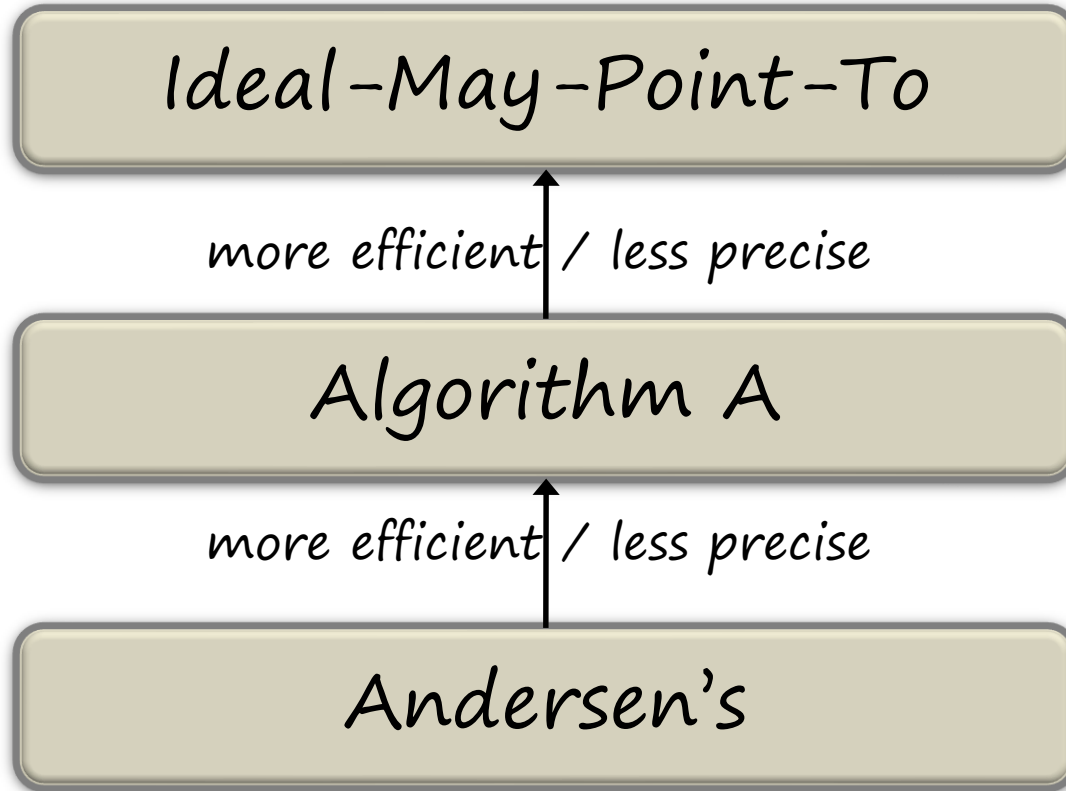


# Andersen's Analysis: An alternative formulation

1. Introduce a constraint variable  $PT_x$  for each program variable  $x$
2. Create a constraint from each assignment statement, as follows:
  - $x = y: PT_x \subseteq PT_y$
  - $*x = y: PT_v \subseteq PT_y, \text{ for all variables } v \text{ in } PT_x$
  - $x = \&y: PT_x \subseteq \{y\}$
  - $x = *y: PT_x \subseteq PT_v, \text{ for all variables } v \text{ in } PT_y$
3. Find least solution to set of all constraints that were generated above. (A solution is a mapping from constraint variables to sets of program variables.) Emit this least solution as the final solution.
  - Note: Solution  $s1$  dominates Solution  $s2$  if for each program variable  $v$ ,  $s2(PT_v) \subseteq s1(PT_v)$

Note: This approach computes exact same result as previous approach that collapses program and then uses Algorithm A.

# May-Point-To Analyses



# Andersen's Analysis: Further Optimizations and Extensions

- Fahndrich et al., Partial online cycle elimination in inclusion constraint graphs, PLDI 1998.
- Rountev and Chandra, Offline variable substitution for scaling points-to analysis, 2000.
- Heintze and Tardieu, Ultra-fast aliasing analysis using CLA: a million lines of C code in a second, PLDI 2001.
- M. Hind, Pointer analysis: Haven't we solved this problem yet?, PASTE 2001.
- Hardekopf and Lin, The ant and the grasshopper: fast and accurate pointer analysis for millions of lines of code, PLDI 2007.
- Hardekopf and Lin, Exploiting pointer and location equivalence to optimize pointer analysis, SAS 2007.
- Hardekopf and Lin, Semi-sparse flow-sensitive pointer analysis, POPL 2009.

# Context-Sensitivity Etc.

- Liang & Harrold, Efficient computation of parameterized pointer information for interprocedural analyses. SAS 2001.
- Lattner et al., Making context-sensitive points-to analysis with heap cloning practical for the real world, PLDI 2007.
- Zhu & Calman, Symbolic pointer analysis revisited. PLDI 2004.
- Whaley & Lam, Cloning-based context-sensitive pointer alias analysis using BDD, PLDI 2004.
- Rountev et al. Points-to analysis for Java using annotated constraints. OOPSLA 2001.
- Milanova et al. Parameterized object sensitivity for points-to and side-effect analyses for Java. ISSTA 2002.

# Applications

- *Compiler optimizations*
- *Verification & Bug Finding*
  - *use in preliminary phases*
  - *use in verification itself*