

# A Theory of Slicing for Probabilistic Control Flow Graphs

Torben Amtoft (Kansas State University)  
Anindya Banerjee (IMDEA Software Institute)

FoSSaCS, April 2016

Setting

Motivating Examples

Correctness Condition

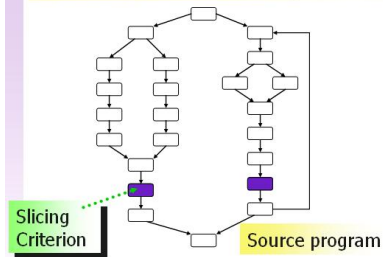
Semantics

Algorithm

Conclusion

# What is Program Slicing?

## Static Backwards Slicing



Pick one or more program points of interest, called the **slicing criterion**

Setting

Motivating Examples

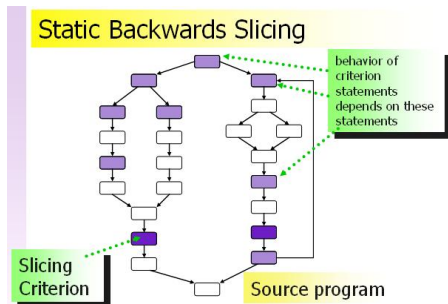
Correctness Condition

Semantics

Algorithm

Conclusion

# What is Program Slicing?



Setting

Motivating Examples

Correctness Condition

Semantics

Algorithm

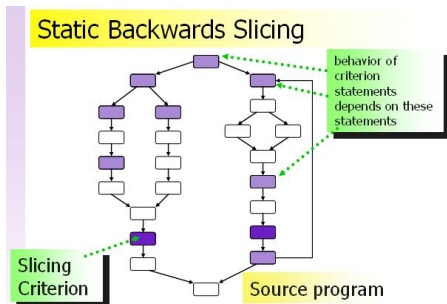
Conclusion

Walk **backwards** to find the nodes (the **slice set**) that the nodes in the slicing criterion **depend** on

- ▶ through **data dependence**, or
- ▶ through **control dependence**

**Remove** nodes **not** in the slice set.

# What is Program Slicing?



Applications include

- ▶ compiler optimizations
- ▶ debugging
- ▶ model checking
- ▶ protocol understanding

Setting

Motivating Examples

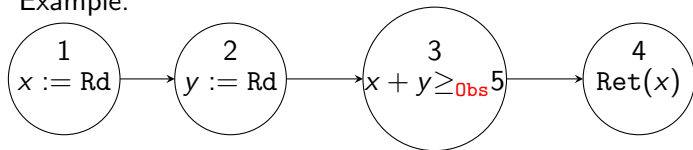
Correctness Condition

Semantics

Algorithm

Conclusion

Example:



We shall work with **CFGs** (control flow graphs) with

- ▶ a unique **End** node, returning the final result
- ▶ **random assignments** from a given distribution
  - ▶ In this talk, we shall always use the one that assigns each of 0,1,2,3 equal probability (0.25)
- ▶ **conditioning** nodes (**Ob**serve) which remove “undesired” value combinations.

**Applications:** see excellent survey article [ICSE'2014] by Andrew Gordon et al

Setting

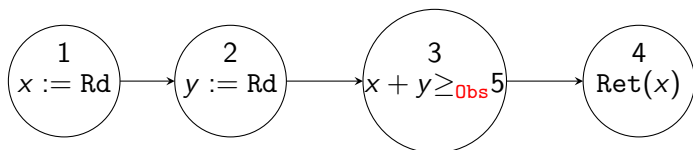
Motivating Examples

Correctness Condition

Semantics

Algorithm

Conclusion



Semantics is expressed using **distributions** which assign **probabilities** to stores.

- ▶ special deterministic case: one store has probability 1, all other stores have probability 0

Distribution  $D$  **before** node 3:

$$D(\{x \mapsto i, y \mapsto j\}) = 1/16 \text{ for } i, j \in 0..3$$

Distribution  $D$  **after** node 3:

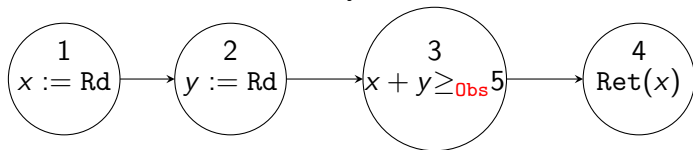
$$D(\{x \mapsto 3, y \mapsto 2\}) = 1/16$$

$$D(\{x \mapsto 2, y \mapsto 1\}) = 0$$

Thus  $\sum D < 1$  is possible (can later be normalized)

# Slicing: the Challenge

**Question:** can we slice away nodes 2 and 3 in



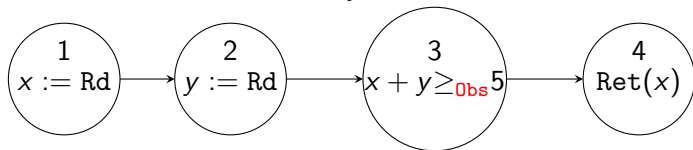
so as to arrive at



In the original program:

# Slicing: the Challenge

**Question:** can we slice away nodes 2 and 3 in



so as to arrive at



In the original program:

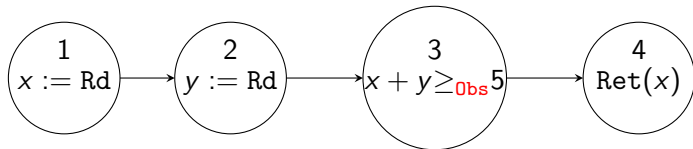
- ▶ classically, 4 depends **only** on 1, so **yes**

[Setting](#)[Motivating Examples](#)[Correctness Condition](#)[Semantics](#)[Algorithm](#)[Conclusion](#)



## Slicing: the Challenge

**Question:** can we slice away nodes 2 and 3 in



so as to arrive at



In the original program:

- ▶ classically, 4 depends **only** on 1, so **yes**
- ▶ but the final distribution of  $x$  is **skewed** ( $x \leq 1$  is **impossible**) so **NO**

We need to be more careful!

- ▶ state what it means for slicing to be **correct** in a probabilistic setting
- ▶ give **syntactic** conditions that **guarantee semantic** correctness
- ▶ present **algorithm** to find **best** syntactic slice.

Setting

Motivating Examples

Correctness Condition

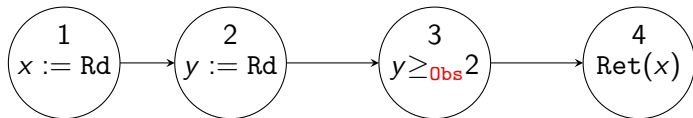
Semantics

Algorithm

Conclusion

## When Slicing is Correct: Semantic Definition

**Question:** can we slice away nodes 2 and 3 in



Final distribution  $D$ :

$$D(\{x \mapsto i, y \mapsto j\}) = 1/16 \text{ for } i \in 0..3, j \in 2..3$$

which restricted to the returned variable  $x$  is

$$D(\{x \mapsto i\}) = 1/8 \text{ for } i \in 0..3$$

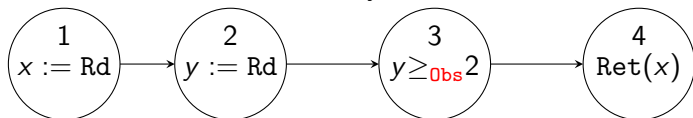
Final distribution  $\Delta$  of sliced program:

$$\Delta(\{x \mapsto i\}) = 1/4 \text{ for } i \in 0..3$$

With  $c = 0.5$  we have  $D = c \cdot \Delta$  and shall therefore say that slicing is **correct** as it does **not skew** the distribution of the **relevant** variable  $x$ .

# Why Slicing is Correct: Syntactic Criterion

**Question:** how to infer, without semantic calculations, that it is correct to slice away nodes 2 and 3 in



The correctness relies on the fact that

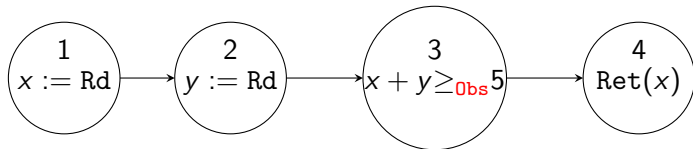
- ▶ at node 4,  $x$  and  $y$  are **probabilistically independent** which will be the case since
- ▶ the set of nodes  $\{1\}$  that may influence  $x$  is **disjoint** from the set of nodes  $\{2\}$  that may influence  $y$ .

**Initial Finding:** A conditioning can be sliced away if

- ▶ the nodes that the End node depends on, and
  - ▶ the nodes that the conditioning depends on
- have nothing in common.

# When Slicing is Incorrect, and Why

We saw it is **incorrect** to slice away nodes 2 and 3 in



since the final distribution  $D$  has say

$$D(\{x \mapsto 1\}) = 0 \text{ but } D(\{x \mapsto 2\}) = 1/16$$

and with  $\Delta$  the uniform distribution of  $x$  in the sliced program we thus for all  $c$  have

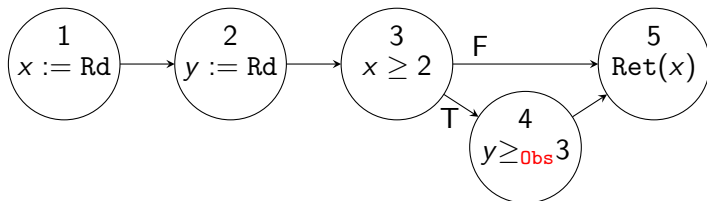
$$D \neq c\Delta$$

And indeed, our tentative syntactic correctness criterion will disallow slicing, since

- ▶ the `End` node (data) depends on node 1, and
- ▶ the conditioning node (data) depends **also** on node 1.

# When Slicing is Incorrect, and Why (II)

Can we slice away nodes 2, 3 and 4 in



Setting

Motivating Examples

Correctness Condition

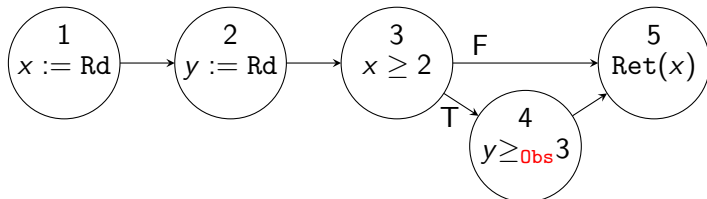
Semantics

Algorithm

Conclusion

# When Slicing is Incorrect, and Why (II)

Can we slice away nodes 2, 3 and 4 in



**No**, since the final distribution  $D$  is skewed:

$$D(\{x \mapsto 1\}) = 1/4 \text{ but } D(\{x \mapsto 2\}) = 1/16$$

And indeed, our tentative syntactic correctness criterion will disallow slicing, since

- ▶ the End node depends on node **1**, and
- ▶ the conditioning node **control** depends on node 3 which data depends on node **1**.

# Tentative Syntactic Correctness Criterion

It appears that for  $Q$  to be a correct slice, we must require that

- ▶  $Q$  is “closed under dependency”
- ▶ all conditioning nodes not in  $Q$  must belong to another set,  $Q_0$ , such that
  - ▶  $Q$  and  $Q_0$  are disjoint
  - ▶  $Q_0$  is also closed under dependency.

We shall soon refine these conditions.

Setting

Motivating Examples

Correctness Condition

Semantics

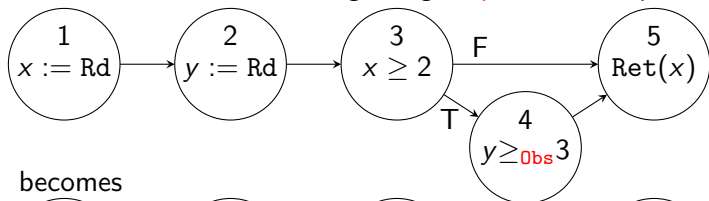
Algorithm

Conclusion

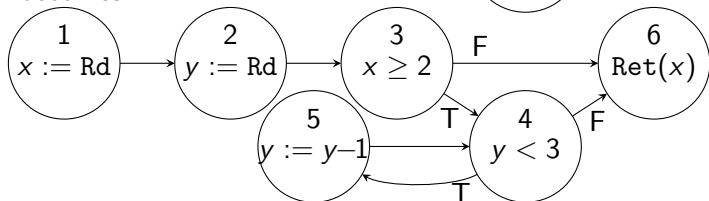


# Why Slicing Away Loops may be Incorrect

We can encode conditioning using **loops**: the example



becomes



Above we can thus **not** slice away all nodes but 1 and 6

- ▶ though node 6 **appears** to depend on node 1 only.

We thus need to **augment correctness criterion**:

- ▶ all loops contain at least one node in  $Q \cup Q_0$ .

We shall now formalize “ $Q$  is closed under dependency”:

- ▶ data dependency (DD): straight-forward
- ▶ control dependency:  
each node has exactly one next  $Q$ -observable

$v'$  is a next  $Q$ -observable of  $v$  (there is at most one) iff

- ▶  $v' \in Q \cup \{\text{End}\}$
- ▶ all paths from  $v$  to a node in  $Q \cup \{\text{End}\}$  contain  $v'$ .

We say that  $Q$  is a weak slice set if

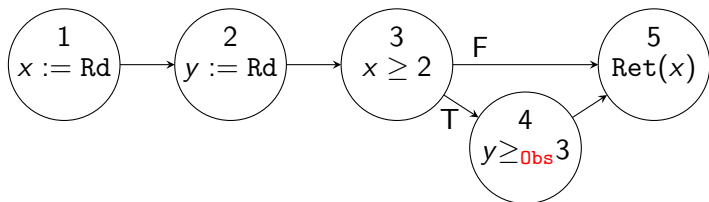
- ▶  $Q$  is closed under DD
- ▶ each node has a next  $Q$ -observable

[Setting](#)[Motivating Examples](#)[Correctness Condition](#)[Semantics](#)[Algorithm](#)[Conclusion](#)

## Weak Slice Sets, Example

Recall:  $v'$  is a next  $Q$ -observable of  $v$  iff

- ▶  $v' \in Q \cup \{\text{End}\}$
- ▶ all paths from  $v$  to a node in  $Q \cup \{\text{End}\}$  contain  $v'$ .



- ▶ With  $Q = \{2, 4\}$ ,  $Q$  is closed under DD but node **3** does **not** have a next  $Q$ -observable:
  - ▶ there is a path from 3 to 5 that does not contain 4
  - ▶ there is a path from 3 to 4 that does not contain 5
- ▶ With  $Q = \{2, 3, 4\}$ , all nodes have next  $Q$ -observables but is **not** closed under DD
- ▶ But  $Q = \{1, 2, 3, 4\}$  is a weak slice set

We say that  $Q$  is a **correct** slice if there exists  $Q_0$  such that  $(Q, Q_0)$  is a **slicing pair** in that

- ▶  $Q$  and  $Q_0$  are weak slice sets
- ▶  $Q \cap Q_0 = \emptyset$
- ▶ all conditioning nodes belong to  $Q \cup Q_0$
- ▶ all loops contain a node in  $Q \cup Q_0$ .

Setting

Motivating Examples

Correctness Condition

Semantics

Algorithm

Conclusion

# Semantics of Probabilistic CFGs

## One-step reduction $(v, D) \rightarrow (v', D')$

- ▶  $v'$  successor of  $v$
- ▶ transforms  $D$  into  $D'$
- ▶ defined for (random) assignments, and conditioning

## Multi-step reduction $(v, D) \Rightarrow (v', D')$

- ▶  $v'$  postdominates  $v$
- ▶ combines paths from  $v$  to  $v'$  that may contain cycles but do not contain  $v'$  until the very end
- ▶ defined as the limit (in  $D'$ ) of an inductively defined relation  $(v, D) \xRightarrow{k} (v', D')$  where  $k$  bounds the number of cycles.

**Conjecture:** for CFGs produced from structured programs, this semantics will coincide with the standard denotational semantics [Kozen, Hur]

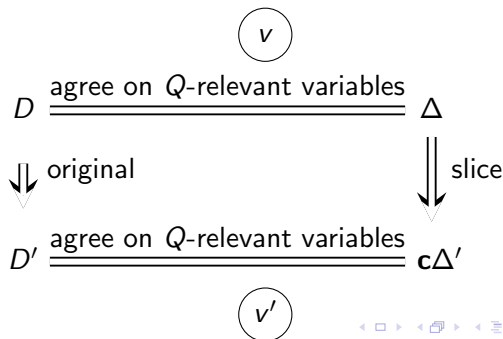
[Setting](#)[Motivating Examples](#)[Correctness Condition](#)[Semantics](#)[Algorithm](#)[Conclusion](#)

# Syntactic Criteria Imply Semantic Correctness

Assume that

- ▶ we have slicing pair  $(Q, Q_0)$
- ▶  $v'$  postdominates  $v$
- ▶ at  $v$ , the  $Q$ -relevant variables and the  $Q_0$ -relevant variables are **probabilistically independent** in  $D$ .

Then there always exists a real number  $c$  with  $0 \leq c \leq 1$  such that the below diagram commutes:



# Probabilistic Independence

When are  $R$  and  $R_0$  **independent** in  $D$ ?

- ▶ **classical** definition: if for all  $s/s_0$  with domain  $R/R_0$ :

$$D(s \oplus s_0) = D(s) \cdot D(s_0)$$

- ▶ in our setting when  $\sum D$  may not equal 1: instead

$$D(s \oplus s_0) \cdot \sum D = D(s) \cdot D(s_0).$$

Let  $(Q, Q_0)$  be a slicing pair, and assume

$(v, D) \xRightarrow{k} (v', D')$ .

- ▶ If at  $v$ , the  $Q$ -relevant variables are independent of the  $Q_0$ -relevant variables in  $D$
- ▶ then at  $v'$ , the  $Q$ -relevant variables are independent of the  $Q_0$ -relevant variables in  $D'$ .

# Algorithm to Compute Best Syntactic Slice

We have an  $O(n^3)$  algorithm to find the **best** slice.

Subcomponents:

1. a function that given  $Q$  either
  - ▶ **confirms** that  $Q$  is a weak slice set, **or**
  - ▶ returns  $C \neq \emptyset$  such that  $C$  is contained in all weak slice sets containing  $Q$

and which works by a **backwards** breadth-first search through nodes **not** in  $Q$

- ▶  $C$  will contain nodes reachable from **two** nodes
2. a function that given  $Q$  finds the **least weak slice set** containing  $Q$

Setting

Motivating Examples

Correctness Condition

Semantics

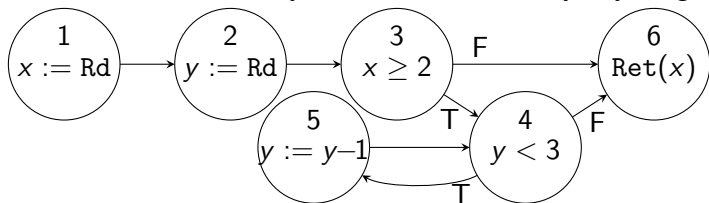
Algorithm

Conclusion

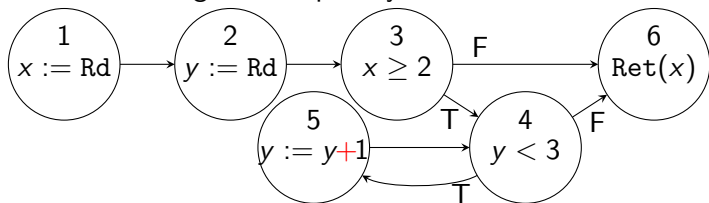


# Semantic vs Syntactic Slices

Recall it is semantically unsound to slice away anything in



but if we change the loop body:



then all but 1 and 6 can be sliced away

- ▶ though  $\{1, 6\}$  does **not** meet our **syntactic** criteria.

Computing the best **semantic** slice is clearly **undecidable**

One may in various ways **transform** the CFG so that the semantics is preserved while smaller slices may be generated:

- ▶ a conditioning  $\text{Observe}(B)$  may be removed if  $B$  can be shown to always hold at that node
- ▶ after a conditioning of the form  $\text{Observe}(x = c)$ , insert an assignment  $x := c$ .

Setting

Motivating Examples

Correctness Condition

Semantics

Algorithm

Conclusion

Our main inspiration is [Hur et al, PLDI'14].

- ▶ they present algorithm for **slicing** of **probabilistic structured** programs
  - ▶ involving various preprocessing
  - ▶ doing various optimizations along the way.
- ▶ conditioning gives rise to a new kind of dependency:
  - ▶ if conditioning depends on  $x$  and  $y$  then any slice that includes  $y$  must also include  $x$
- ▶ **no separation** between **specification** and **implementation**
  - ▶ this makes correctness proof more complex
- ▶ no analysis of asymptotic running time

Setting

Motivating Examples

Correctness Condition

Semantics

Algorithm

Conclusion

- ▶ framework for slicing of probabilistic programs which separates specification from implementation
- ▶ extends in non-trivial way classical slicing frameworks (as generalized by Danicic [TCS 2011])
- ▶ presents operational semantics for probabilistic CFGs
- ▶ presents cubic-time algorithm for finding best (syntactic) slice

Future/present work:

- ▶ allow to slice away loops that are know to always (with probability 1) terminate

Setting

Motivating Examples

Correctness Condition

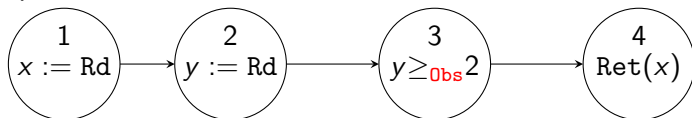
Semantics

Algorithm

Conclusion

# When Algorithm Computes Non-Trivial Slice

We now show how the algorithm finds the best slicing pair for



1. Compute least weak slice set that contains 4:
  - 1.1 close under DD:  $\{1, 4\}$
  - 1.2 nothing more is needed for unique next observable
2. Compute least weak slice set that contains 3 (the conditioning):
  - 2.1 close under DD:  $\{2, 3\}$
  - 2.2 nothing more is needed for unique next observable

As  $\{1, 4\} \cap \{2, 3\}$  are disjoint, this shows that  $Q = \{1, 4\}$  is a valid slice (with  $Q_0 = \{2, 3\}$ ).

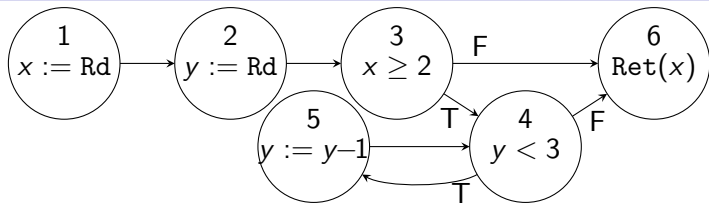
Recall that we require

*for each loop, at least one node is in  $Q \cup Q_0$*

- ▶ this may **appear** to allow **in**compatible solutions
- ▶ but is equivalent to the below

*for each loop,*

*the node(s) with “minimal height” is in  $Q \cup Q_0$*



1. Compute least weak slice set that contains 6:
  - 1.1 close under DD:  $\{1, 6\}$
  - 1.2 nothing more is needed for unique next observable
2. Compute least weak slice set that contains 4 (the loop node closest to 6):
  - 2.1 close under DD:  $\{2, 4, 5\}$
  - 2.2 a backwards search from  $\{2, 4, 5\} \cup \{6\}$  will hit 3 from 4 and from 6 so we need to add 3:  $\{2, 3, 4, 5\}$
  - 2.3 again close under DD:  $\{1, 2, 3, 4, 5\}$

As the two results are **not** disjoint, we cannot put the second result in  $Q_0$ ; instead, we must put it in  $Q$  and end up with a trivial slice:  $Q = \{1, 2, 3, 4, 5, 6\}$