

Automatically Generating Features for Learning Program Analysis Heuristics for C-Like Languages

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Static Analysis

- Diverse engineering decisions in static analysis:
 - Context-sensitivity for which procedures?
 - Relational analysis for which variables?
 - Unsoundness for which part of the program?
 - etc.

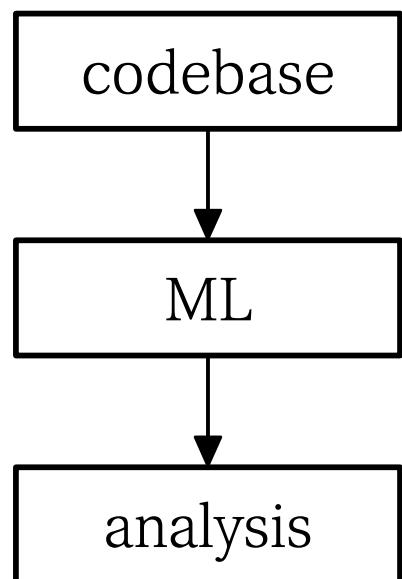
Static Analysis

Data-Driven

- Diverse engineering decisions in static analysis:
 - Context-sensitivity for which procedures?
 - Relational analysis for which variables?
 - Unsoundness for which part of the program?
 - etc.
- Data-driven static analysis aims at **automatically learning analysis heuristics** from the **codebase**.
 - ✓ context-sensitivity heuristics
 - ✓ relation tracking heuristics
 - ✓ unsoundness heuristics
 - ✓ etc.

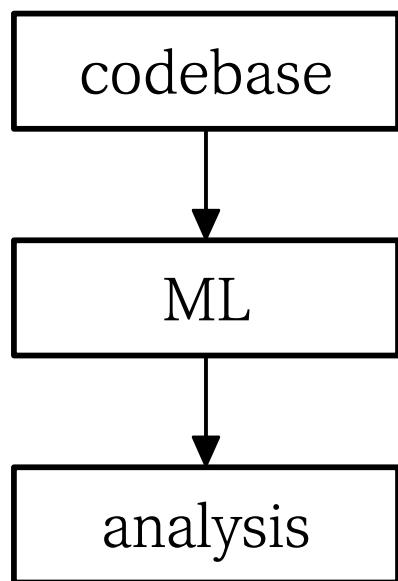
Main Obstacle: Manual Feature Engineering

What people expect

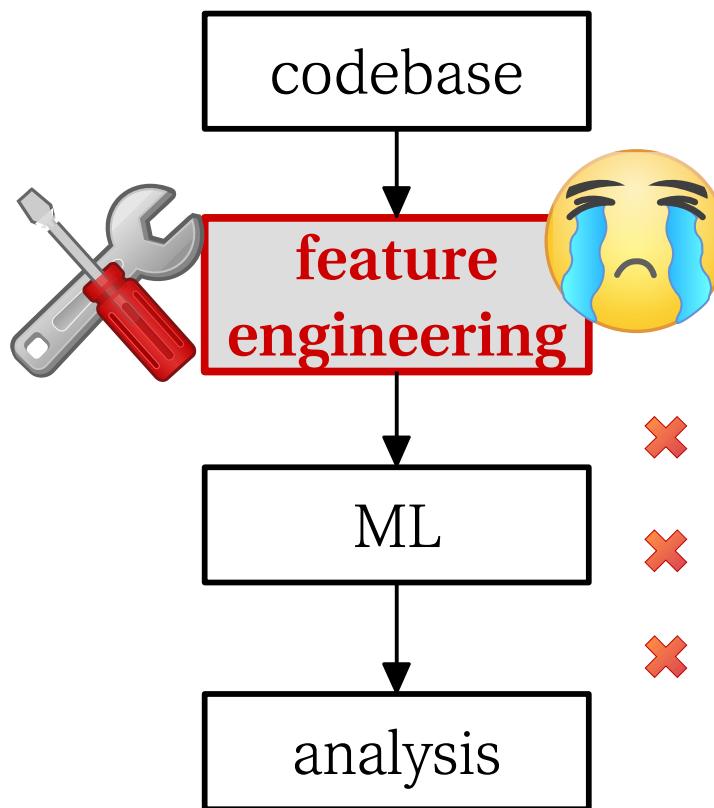


Main Obstacle: Manual Feature Engineering

What people expect



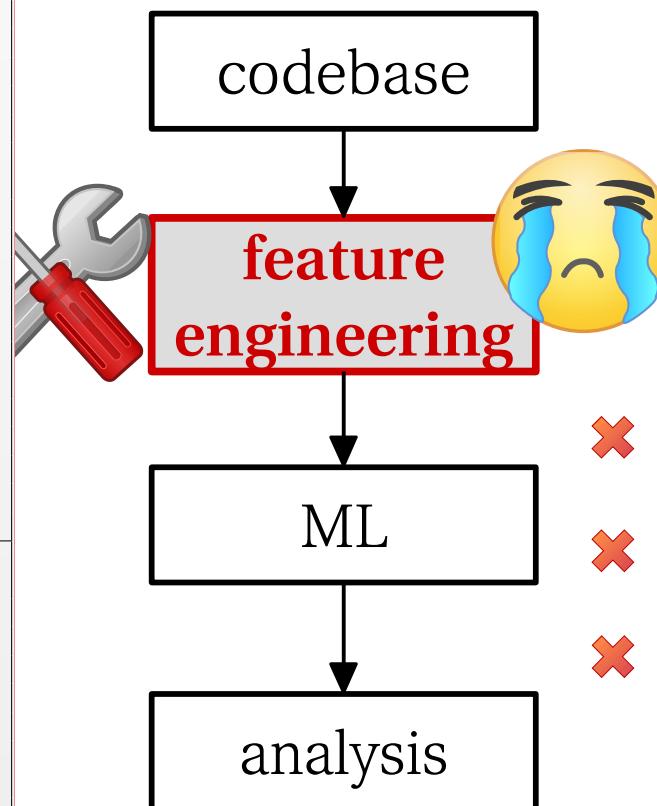
Reality



- ✖ Manual, time-consuming
- ✖ Need for domain expertise
- ✖ Not interchangeable
among different analyses

Main Obstacle: Feature Engineering

Reality

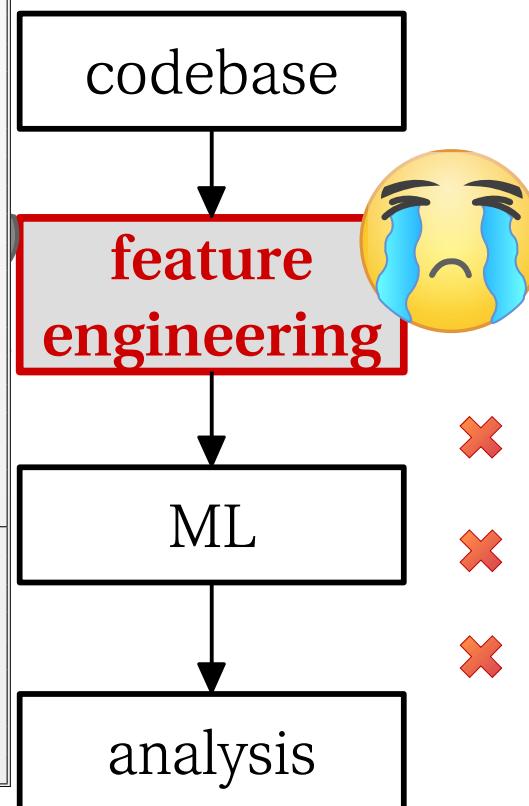


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Type	#	Features
A	1	local variable
	2	global variable
	3	structure field
	4	location created by dynamic memory allocation
	5	defined at one program point
	6	location potentially generated in library code
	7	assigned a constant expression (e.g., $x = c1 + c2$)
	8	compared with a constant expression (e.g., $x < c$)
	9	compared with an other variable (e.g., $x < y$)
	10	negated in a conditional expression (e.g., if ($\neg x$))
	11	directly used in malloc (e.g., malloc(x))
	12	indirectly used in malloc (e.g., $y = x$; malloc(y))
	13	directly used in realloc (e.g., realloc(x))
	14	indirectly used in realloc (e.g., $y = x$; realloc(y))
	15	directly returned from malloc (e.g., $x = \text{malloc}(e)$)
	16	indirectly returned from malloc
	17	directly returned from realloc (e.g., $x = \text{realloc}(e)$)
	18	indirectly returned from realloc
	19	incremented by one (e.g., $x = x + 1$)
	20	incremented by a constant expr. (e.g., $x = x + (1+2)$)
	21	incremented by a variable (e.g., $x = x + y$)
	22	decremented by one (e.g., $x = x - 1$)
	23	decremented by a constant expr (e.g., $x = x - (1+2)$)
	24	decremented by a variable (e.g., $x = x - y$)
	25	multiplied by a constant (e.g., $x = x * 2$)
	26	multiplied by a variable (e.g., $x = x * y$)
	27	incremented pointer (e.g., $p++$)
	28	used as an array index (e.g., $a[x]$)
	29	used in an array expr. (e.g., $x[e]$)
	30	returned from an unknown library function
	31	modified inside a recursive function
	32	modified inside a local loop
	33	read inside a local loop
B	34	$1 \wedge 8 \wedge (11 \vee 12)$
	35	$2 \wedge 8 \wedge (11 \vee 12)$
	36	$1 \wedge (11 \vee 12) \wedge (19 \vee 20)$
	37	$2 \wedge (11 \vee 12) \wedge (19 \vee 20)$
	38	$1 \wedge (11 \vee 12) \wedge (15 \vee 16)$
	39	$2 \wedge (11 \vee 12) \wedge (15 \vee 16)$
	40	$(11 \vee 12) \wedge 29$
	41	$(15 \vee 16) \wedge 29$
	42	$1 \wedge (19 \vee 20) \wedge 33$
	43	$2 \wedge (19 \vee 20) \wedge 33$
	44	$1 \wedge (19 \vee 20) \wedge \neg 33$
	45	$2 \wedge (19 \vee 20) \wedge \neg 33$

Main Obstacle: Feature Engineering

Reality



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Type	#	Features
A	1	local variable
	2	global variable
	3	structure
	4	location
	5	define
	6	location
	7	assign
	8	comp
	9	comp
	10	negate
	11	direct
	12	indirect
	13	direct
	14	indirect
	15	direct
	16	indirect
	17	direct
	18	indirect
	19	increment
	20	increment
	21	increment
	22	decrement
	23	decrement
	24	decrement
	25	multi
	26	multi
	27	increment
	28	used as
	29	used in
	30	return
	31	modification
	32	modification
	33	read in
B	30	$2 \wedge (21 \vee 22) \wedge (14 \vee 15)$
	31	$2 \wedge (21 \vee 22) \wedge \neg(14 \vee 15)$
	32	$2 \wedge 23 \wedge (14 \vee 15)$
	33	$2 \wedge 23 \wedge \neg(14 \vee 15)$
	34	$2 \wedge (21 \vee 22) \wedge (16 \vee 17)$
	35	$2 \wedge (21 \vee 22) \wedge \neg(16 \vee 17)$
	36	$2 \wedge 23 \wedge (16 \vee 17)$
	37	$2 \wedge 23 \wedge \neg(16 \vee 17)$
	38	$(21 \vee 22) \wedge \neg 23$
	39	$(11 \vee 15) \wedge \neg 23$
	40	$2 \wedge (19 \vee 20) \wedge 33$
	41	$1 \wedge (19 \vee 20) \wedge \neg 33$
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Main Obstacle:

widening threshold
(APLAS'16)

Feature Engineering

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	5	define	
	6	location	
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	8	comp	
	9	comp	
	10	negate	
	11	direct	
	12	indire	
	13	direct	
	14	indire	
	15	direct	
	16	indire	
	17	direct	
	18	indire	
	19	incre	
	20	incre	
	21	incre	
	22	decre	
	23	decre	
	24	decre	
	25	multip	
	26	multip	
	27	incre	
	28	used a	
	29	used i	
	30	return	
	31	modif	
	32	modif	
	33	read i	
A	1	Type	#
	2	leaf function	
	3	function containing malloc	
	4	function containing realloc	
	5	function c	
	6	function c	
	7	function c	
	8	function u	
	9	write to a	
	10	read a glo	
	11	write to a	
	12	read from	
	13	directly re	
	14	indirectly re	
	15	directly re	
	16	indirectly re	
	17	indirectly re	
	18	directly re	
	19	indirectly re	
	20	directly re	
	21	indirectly re	
	22	return exp	
	23	return val	
	24	return void	
	25	directly in	
	26	constant i	
	27	invoked with an unknown value	
	28	functions having no arguments	
	29	functions having one argument	
	30	functions having more than one argument	
	31	functions having an integer argument	
	32	functions having a pointer argument	
	33	functions having a structure as an argument	
B	30	2 \wedge (21 \vee 22) \wedge (14 \vee 15)	
	31	2 \wedge (21 \vee 22) \wedge \neg (14 \vee 15)	
	32	2 \wedge 23 \wedge (14 \vee 15)	
	33	2 \wedge 23 \wedge \neg (14 \vee 15)	
	34	2 \wedge (21 \vee 22) \wedge (16 \vee 17)	
	35	2 \wedge (21 \vee 22) \wedge \neg (16 \vee 17)	
	36	2 \wedge 23 \wedge (16 \vee 17)	
	37	2 \wedge 23 \wedge \neg (16 \vee 17)	
	38	(11 \vee 15) \wedge 23	
	39	2 \wedge (19 \vee 20) \wedge 33	
	40	1 \wedge (19 \vee 20) \wedge \neg 33	
	41	2 \wedge (19 \vee 20) \wedge \neg 33	
	42	1 \wedge (19 \vee 20) \wedge 33	
	43	2 \wedge (19 \vee 20) \wedge 33	
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engineering

ML

analysis

- ✖ Manual, time-consuming
- ✖ Need for domain expertise
- ✖ Not interchangeable among different analyses

Main Obstacle: Feature Engineering

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	14	indire	
	15	direct	
	16	indire	
	17	direct	
	18	indire	
	19	incre	
	20	incre	
	21	incre	
	22	decre	
	23	decre	
	24	decre	
	25	multi	
	26	multi	
	27	incre	
	28	used a	
	29	used i	
	30	return	
	31	modif	
	32	modif	
	33	read i	
B	30	2 \wedge (21 \vee 22) \wedge (14 \vee 15)	
	31	2 \wedge (21 \vee 22) \wedge \neg (14 \vee 15)	
	32	2 \wedge 23 \wedge (14 \vee 15)	
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	34	2 \wedge (21 \vee 22) \wedge (16 \vee 17)	
	35	2 \wedge (21 \vee 22) \wedge \neg (16 \vee 17)	
	36	2 \wedge (11 \vee 15) \wedge 23	
	37	2 \wedge (15 \vee 11) \wedge 23	
	38	2 \wedge (21 \vee 22) \wedge 23	
	39	2 \wedge (21 \vee 22) \wedge \neg 23	
	40	2 \wedge (19 \vee 20) \wedge 33	
	41	1 \wedge (19 \vee 20) \wedge \neg 33	
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, time-consuming
 or domain expertise
 interchangeable
 different analyses

Main Obstacle: Feature Engineering

unsoundness
(ICSE'17)

Type	#	Features					Feature Engineering				
A	1	local variable					Description				
	2	global variable					Description				
	3	struct					Description				
	4	location					Description				
	5	define					Description				
	6	location					Description				
	7	assign					Description				
	8	comp					Description				
	9	comp					Description				
	10	negate					Description				
	11	direct					Description				
	12	indirect					Description				
	13	direct					Description				
	14	indirect					Description				
	15	direct					Description				
	16	indirect					Description				
	17	direct					Description				
	18	indirect					Description				
	19	increasing					Description				
	20	increasing					Description				
	21	increasing					Description				
	22	decreasing					Description				
	23	decreasing					Description				
	24	decreasing					Description				
	25	multi					Description				
	26	multi					Description				
	27	increasing					Description				
	28	used as argument					Description				
	29	used as input					Description				
	30	return					Description				
	31	modification					Description				
	32	modification					Description				
	33	read integer					Description				
B	34	B					Description				
	35	1 ∧ 8					Description				
	36	2 ∧ 8					Description				
	37	1 ∧ (Description				
	38	2 ∧ (Description				
	39	1 ∧ (Description				
	40	2 ∧ (Description				
	41	(11 ∨					Description				
	42	(15 ∨					Description				
	43	1 ∧ (19 ∨ 20) ∧ 33					Description				
	44	1 ∧ (19 ∨ 20) ∧ ¬33					Description				
	45	2 ∧ (19 ∨ 20) ∧ ¬33					Description				

Main Obstacle:

context-sensitivity (previous talk)

Feature Engineering

Main Obstacle:

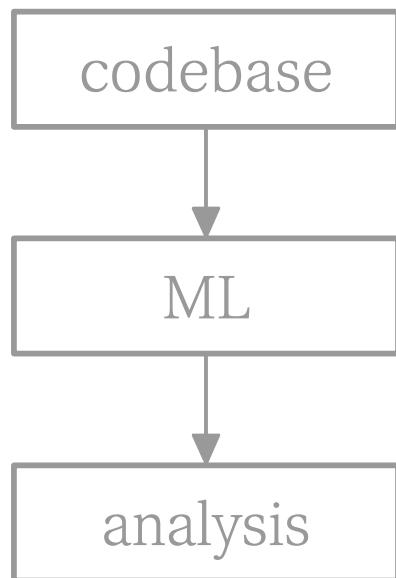
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Feature Engineering

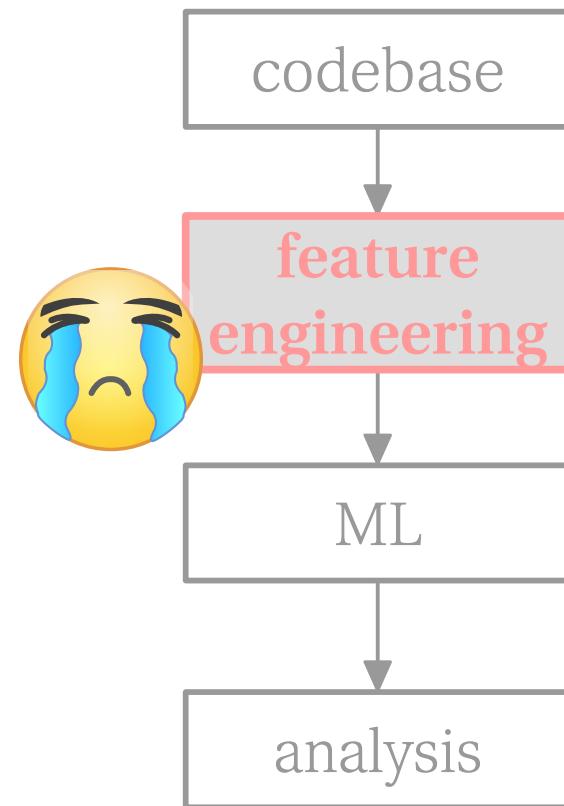
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	19	direct	
	20	indire	
	21	direct	
	22	indire	
	23	direct	
	24	indire	
	25	direct	
	26	indire	
	27	direct	
	28	indire	
	29	direct	
	30	indire	
	31	direct	
	32	indire	
	33	read i	
	34	used i	
	35	used i	
	36	return	
	37	modif	
	38	modif	
	39	read i	
	40	used i	
	41	used i	
	42	return	
	43	modif	
	44	modif	
	45	read i	
	46	used i	
	47	used i	
	48	return	
	49	modif	
	50	modif	
	51	read i	
	52	used i	
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	54	return	
	55	modif	
	56	modif	
	57	read i	
	58	used i	
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	60	return	
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	63	read i	
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	66	return	
	67	modif	
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	69	read i	
	70	used i	
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	72	return	
	73	modif	
	74	modif	
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	78	return	
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	84	return	
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	243	read i	
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	246	return	
	247	modif	
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	249	read i	
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	252	return	
	253	modif	
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We Aim At Generating Features Automatically

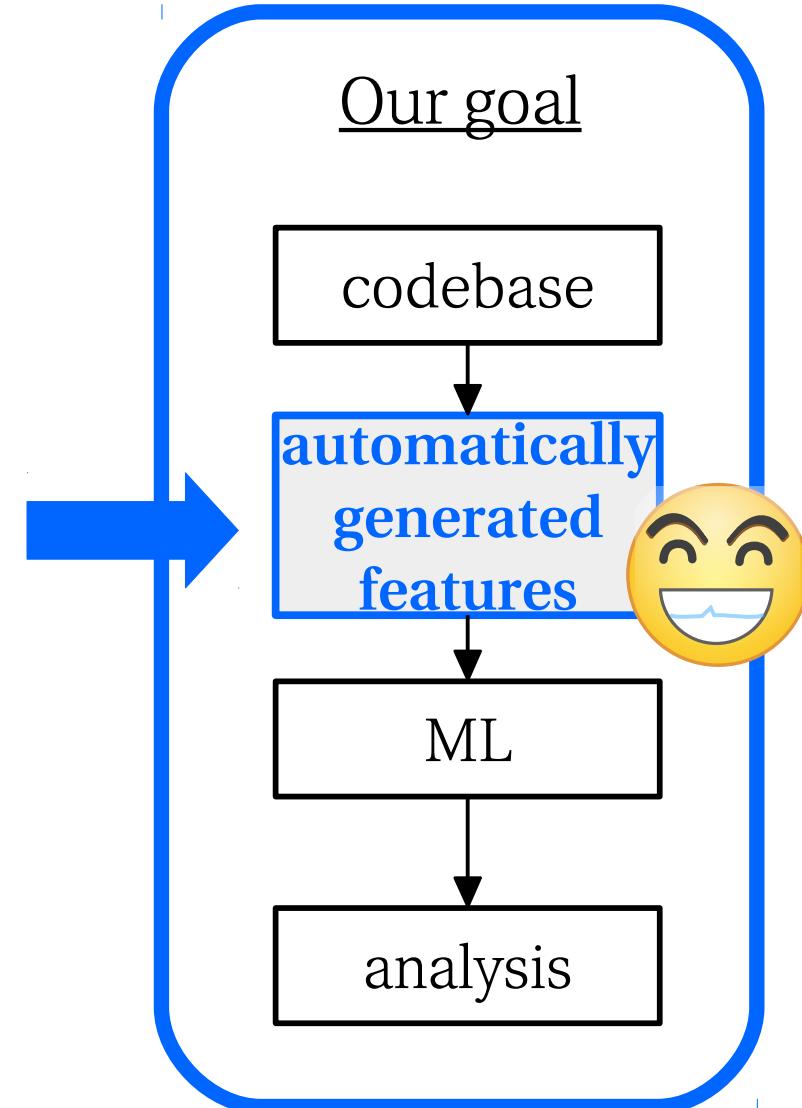
What people expect



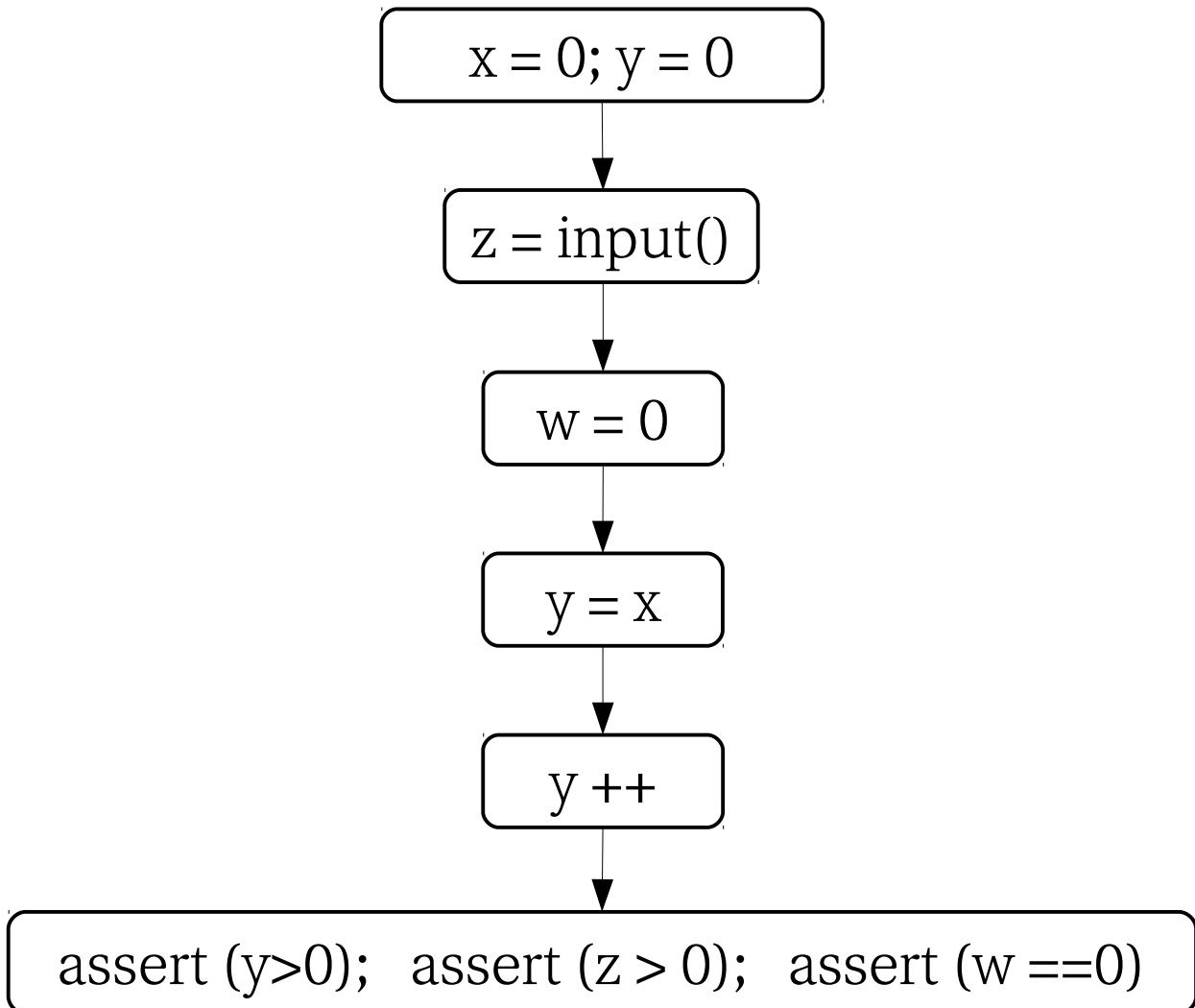
Reality



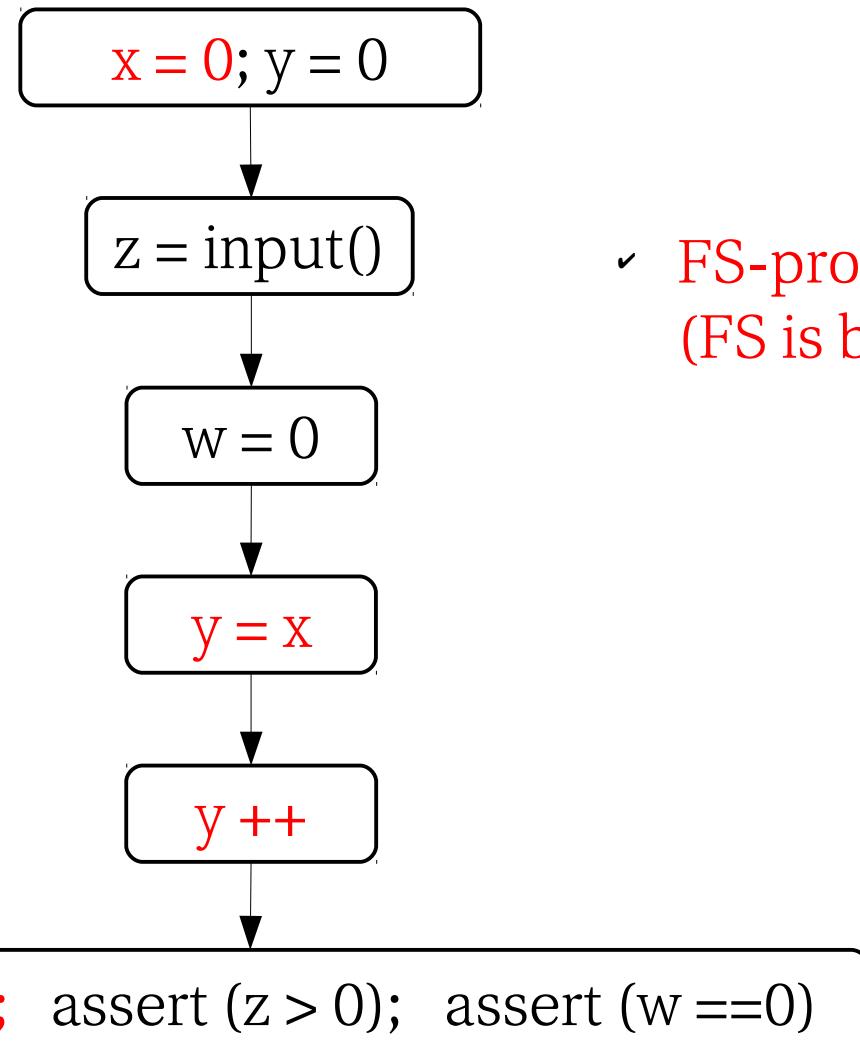
Our goal



Example: Partially Flow-Sensitive Interval Analysis

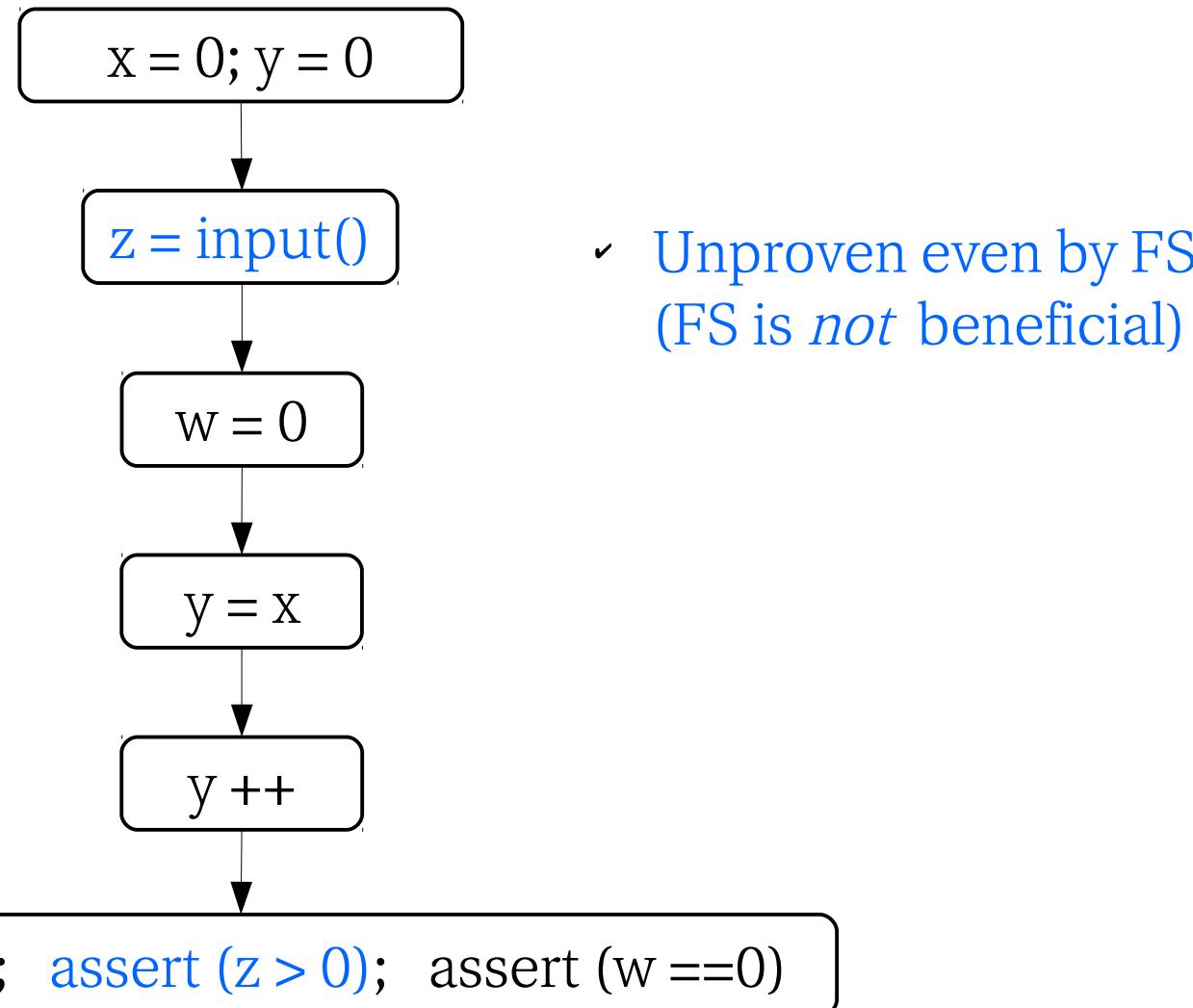


Example: Partially Flow-Sensitive Interval Analysis

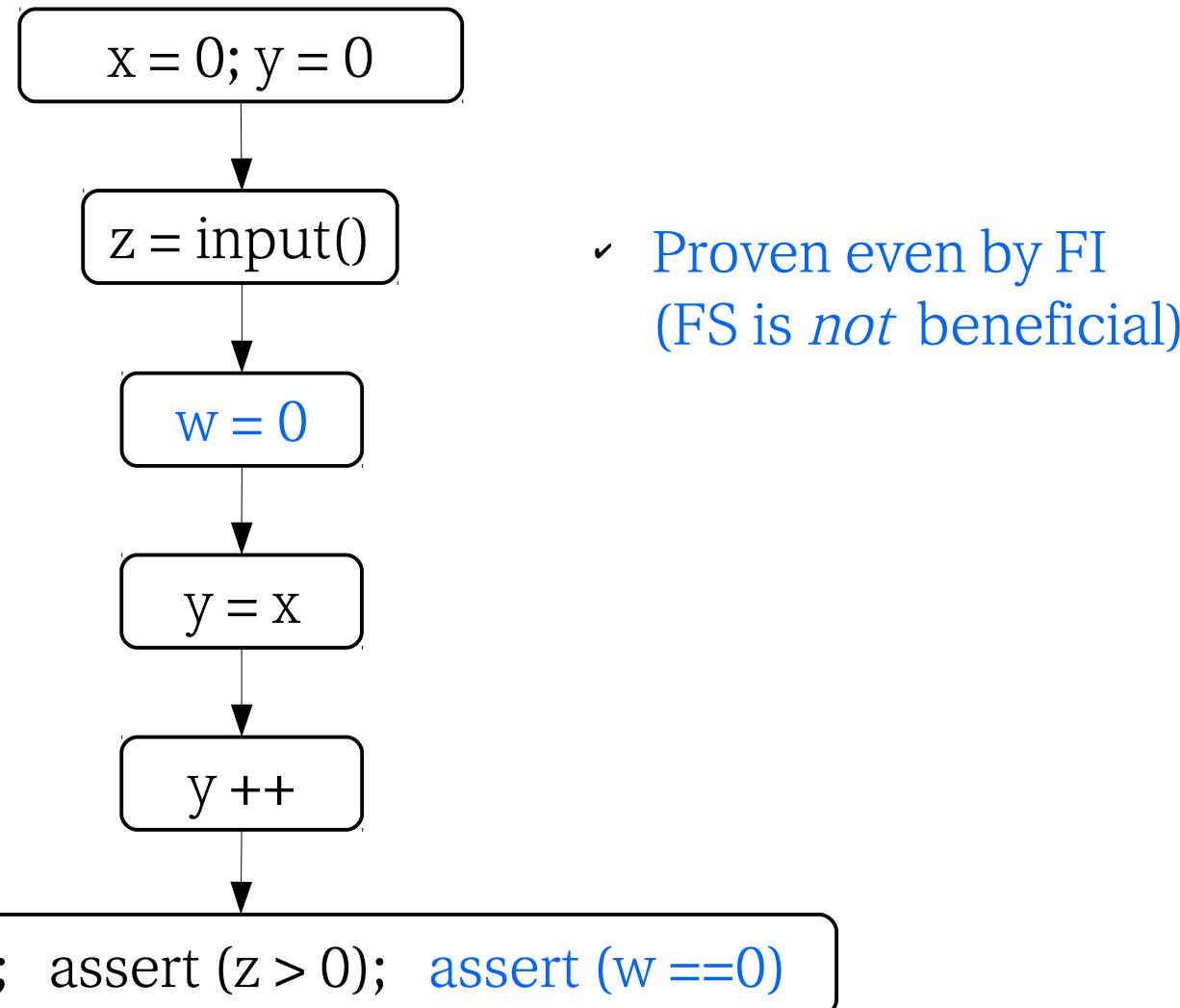


- ✓ FS-proven but FI-unproven
(FS is beneficial)

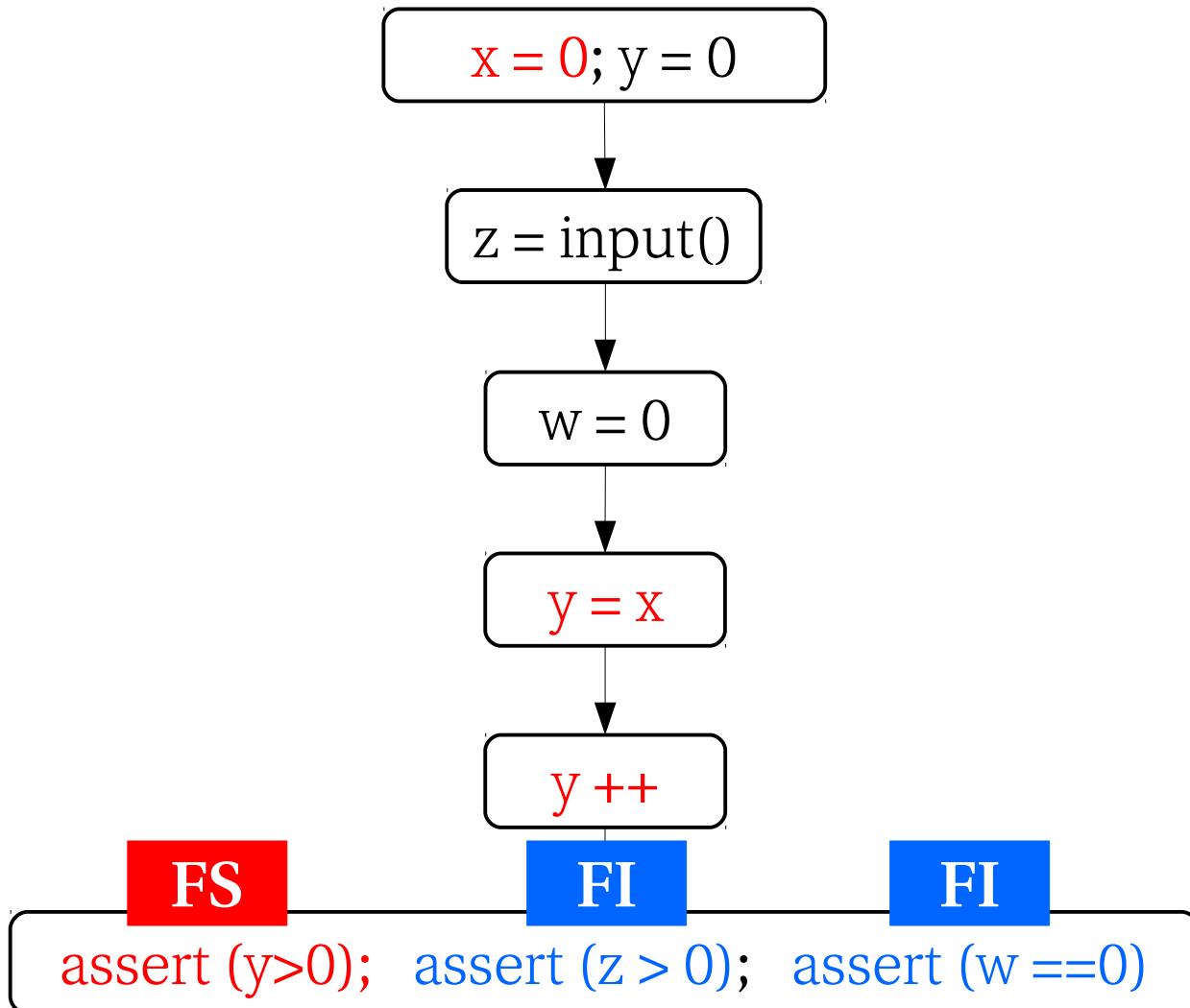
Example: Partially Flow-Sensitive Interval Analysis



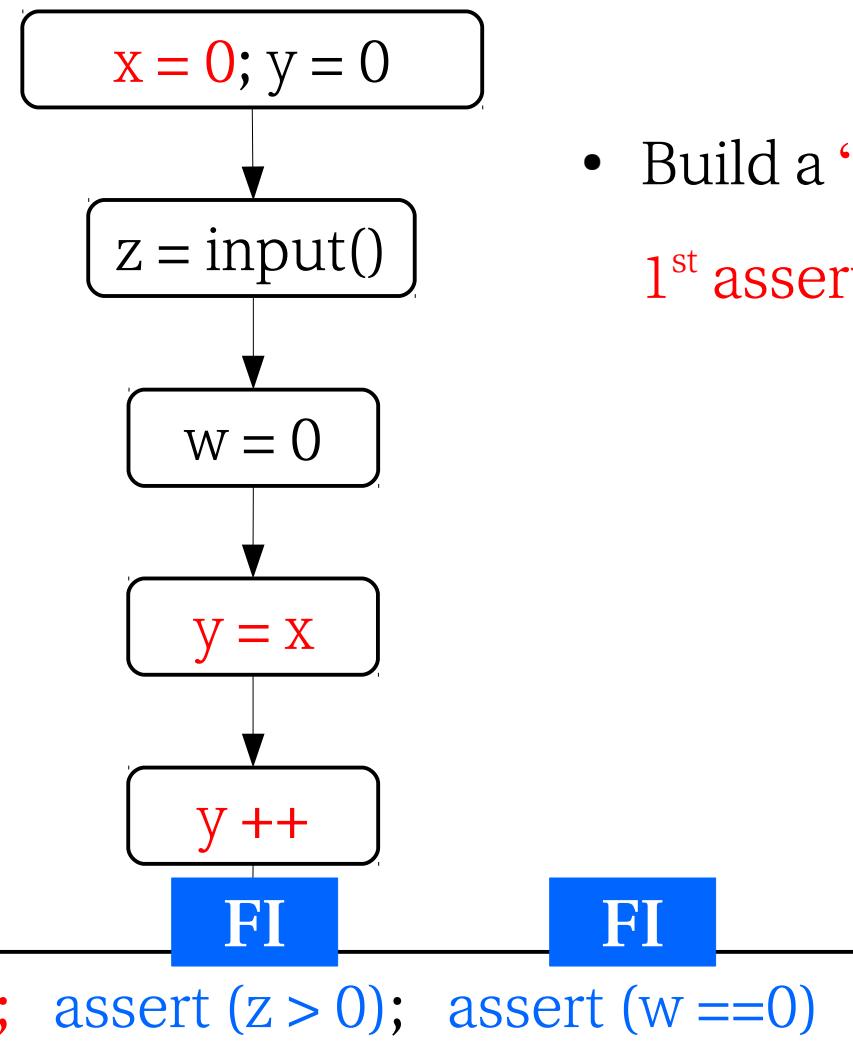
Example: Partially Flow-Sensitive Interval Analysis



Example: Partially Flow-Sensitive Interval Analysis



Example: Partially Flow-Sensitive Interval Analysis



- Build a “classifier” that selects the 1st assertion only.

Building a Classifier

- Usual procedure
 - (1) Design a good set of features manually.

$$F = \{f_1, \dots, f_k\}$$

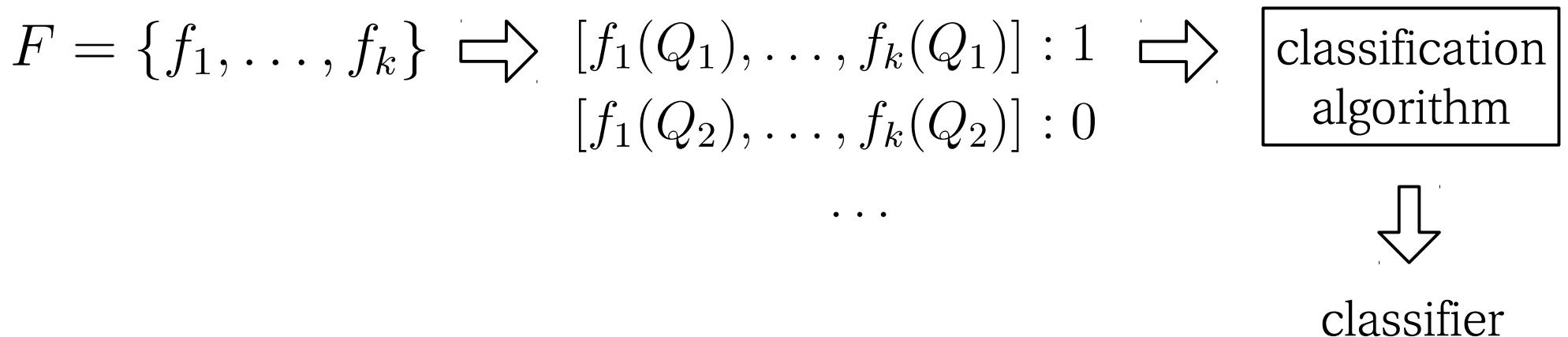
Building a Classifier

- Usual procedure
 - (1) Design a good set of features manually.
 - (2) Generate labeled data.

$$F = \{f_1, \dots, f_k\} \Rightarrow \begin{aligned}[f_1(Q_1), \dots, f_k(Q_1)] &: 1 \\ [f_1(Q_2), \dots, f_k(Q_2)] &: 0 \\ &\dots\end{aligned}$$

Building a Classifier

- Usual procedure
 - (1) Design a good set of features manually.
 - (2) Generate labeled data.
 - (3) Run an off-the-shelf classification algorithm.



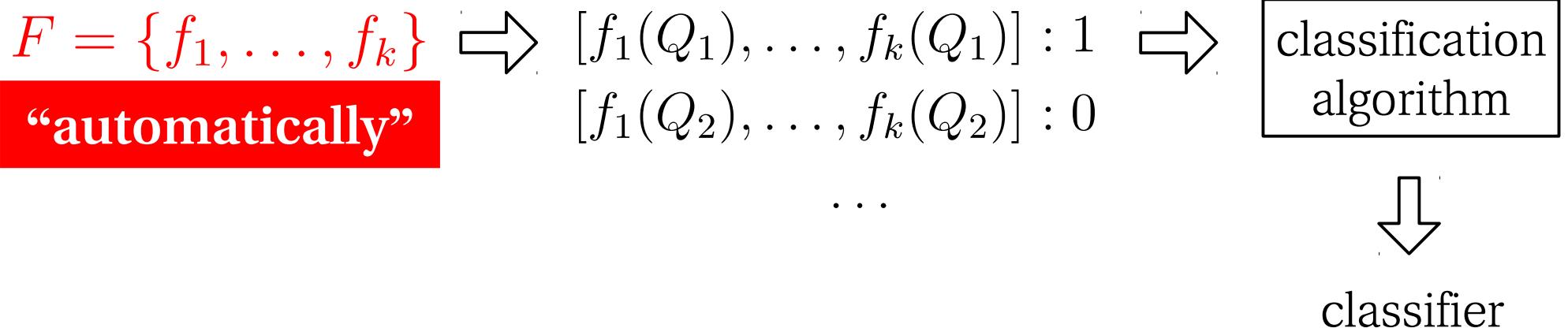
Building a Classifier

automatically

Our

- Usual procedure

- (1) Design a good set of features manually.
- (2) Generate labeled data.
- (3) Run an off-the-shelf classification algorithm.



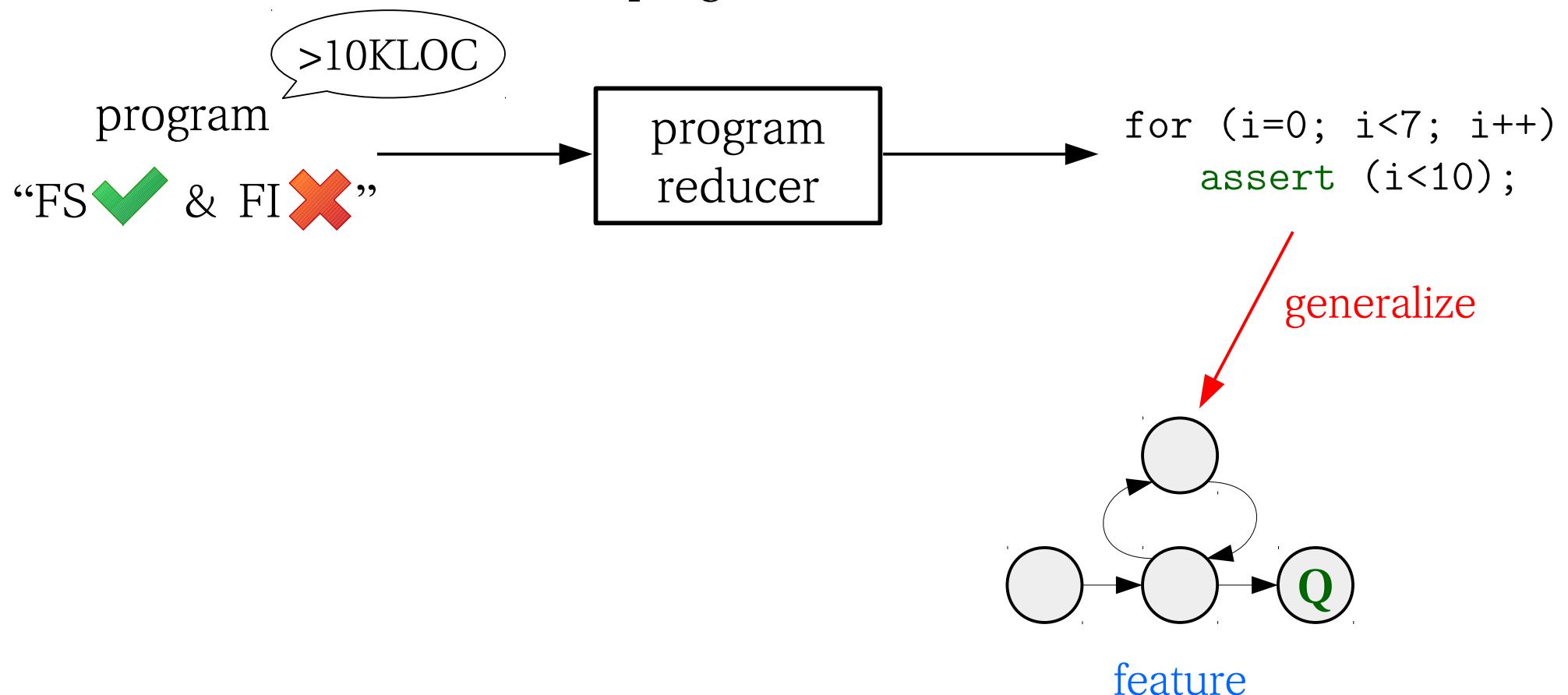
Highlight: Key Ideas

1. Capture the key reason why FS is beneficial using a **program reducer**.



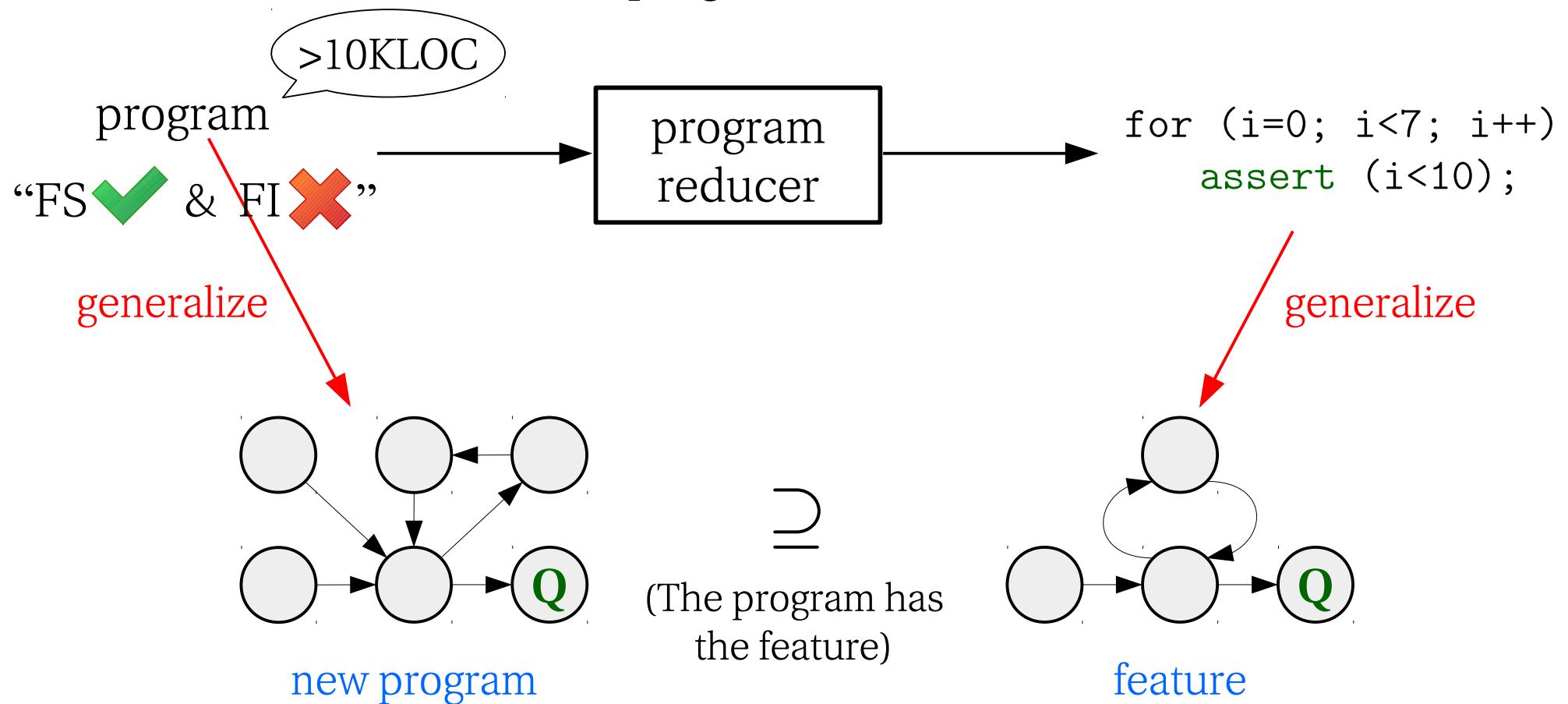
Highlight: Key Ideas

1. Capture the key reason why FS is beneficial using a program reducer.
2. Generalize the reduced program.



Highlight: Key Ideas

1. Capture the key reason why FS is beneficial using a program reducer.
2. Generalize the reduced program.



Highlight: Results

- Generated 38 (interval), 45 (pointer), 44 (Octagon) features.
- Analysis heuristics built on top of automatically generated features
- Excellent balance between cost and precision, e.g.,
 - Partially flow-sensitive interval analysis:

Precision

80.2 %



FI (0)



FS(100)

Cost



FI (1x)

2.0x



FS (46x)

Automatic Feature Generation

Recipe

- (1) Capture the key reasons from the codebase
(using a program reducer).
- (2) Properly generalize the key reasons
(to build generic features).

The end.

(1) Capture The Key Reasons

```
1 a = 0; b = 0;
2 while (1) {
3     b = unknown();
4     if (a > b)
5         if (a < 3)
6             assert (a < 5);
7     a++;
8 }
```

(1) Capture The Key Reasons

```
1 a = 0; b = 0;  
2 while (1) {  
3     b = unknown();  
4     if (a > b)  
5         if (a < 3)  
6             assert (a < 5);  
7     a++;  
8 }
```

reduce →

```
1 a = 0;  
2 while (1) {  
3     if (a < 3)  
4         assert (a < 5);  
5     a++;  
6 }
```

(FS: “ $a < 3$ before assertion in the loop”)

e.g., C-Reduce

- Use a program reducer to generate a **feature program**.
- The reduction preserves an invariant ϕ :

$$\phi(p, q) \equiv FI(p, q) = \text{unproven} \wedge FS(p, q) = \text{proven}$$

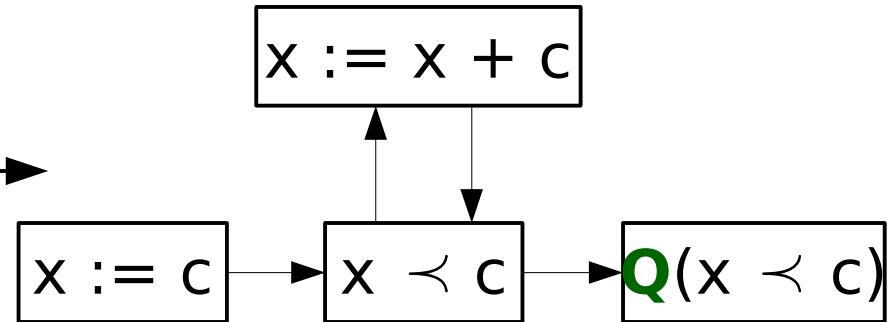
(2) Generalize The Key Reasons

```
1  a = 0;
2  while (1) {
3      if (a < 3)
4          assert (a < 5);
5      a++;
6 }
```

(2) Generalize The Key Reasons

```
1  a = 0;  
2  while (1) {  
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4          assert (a < 5);  
5      a++;  
6 }
```

abstract

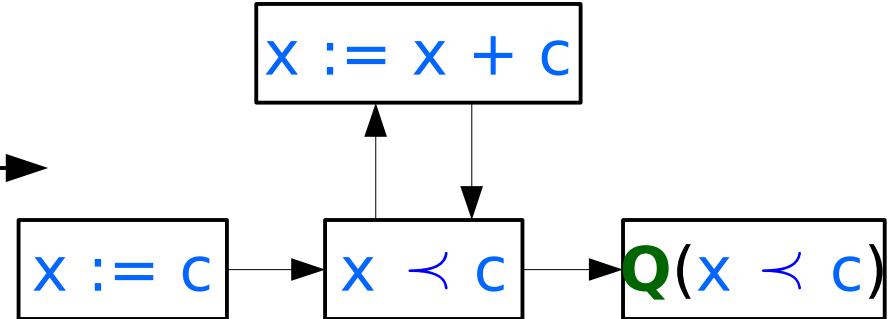


- Properly generalize the feature program to an **abstract data-flow graph (= feature)**.

(2) Generalize The Key Reasons

```
1  a = 0;  
2  while (1) {  
3      if (a < 3)  
4          assert (a < 5);  
5      a++;  
6 }
```

abstract



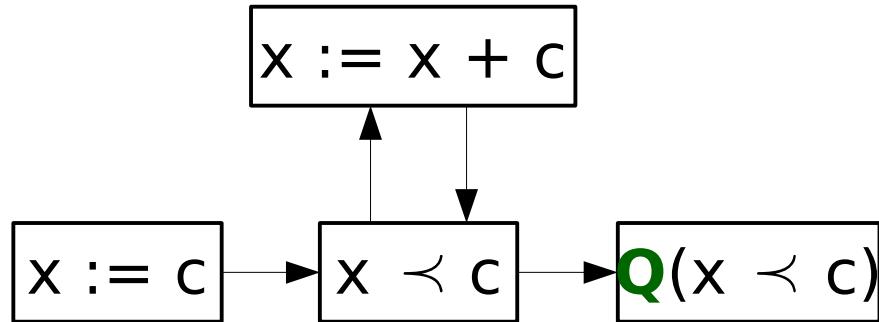
- Properly generalize the feature program to an **abstract data-flow graph (= feature)**.
- The **right level of abstraction** is automatically identified by an iterative search and cross validation.

Generalization **vs.** Preservation

Feature Check = Graph Inclusion Check

original program:

feature:

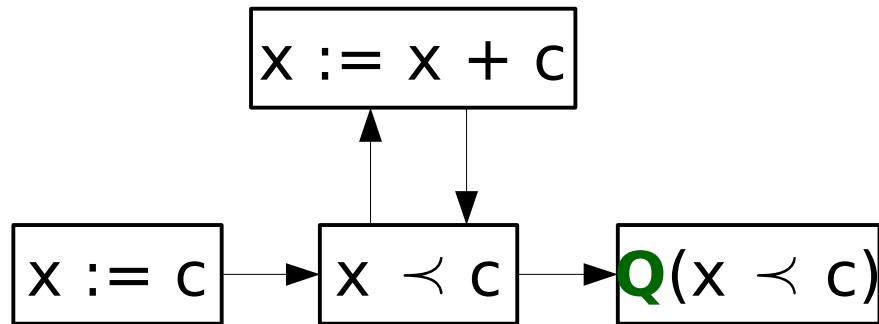


```
1  a = 0; b = 0;
2  while (1) {
3      b = unknown();
4      if (a > b)
5          if (a < 3)
6              assert (a < 5);
7      a++;
8 }
```

Feature Check = Graph Inclusion Check

original program:

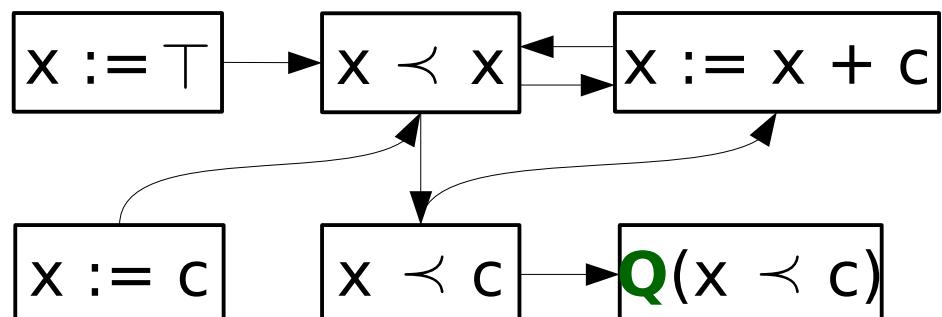
feature:



```
1  a = 0; b = 0;  
2  while (1) {  
3      b = unknown();  
4      if (a > b)  
5          if (a < 3)  
6              assert (a < 5);  
7      a++;  
8 }
```

A blue arrow points from the original program code down to the abstract data-flow graph, indicating the mapping between the two representations.

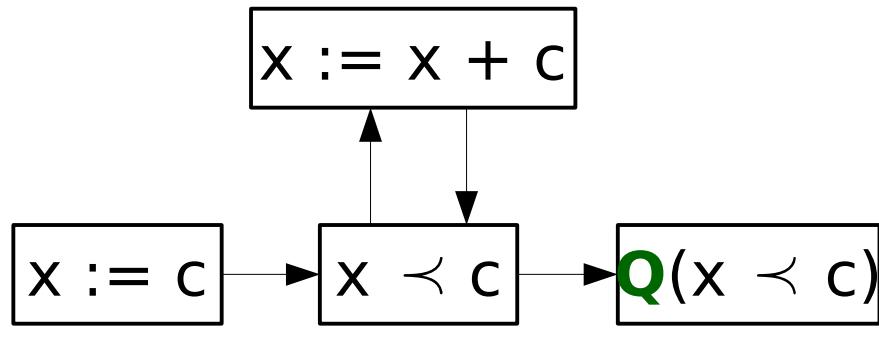
abstract data-flow graph:



Feature Check = Graph Inclusion Check

original program:

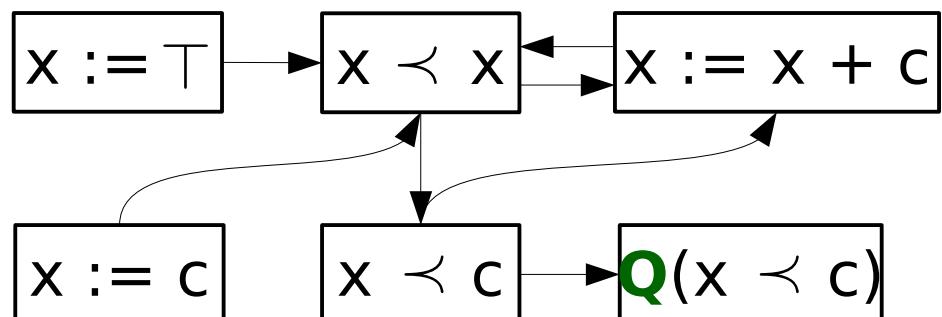
feature:



in

```
1 a = 0; b = 0;
2 while (1) {
3     b = unknown();
4     if (a > b)
5         if (a < 3)
6             assert (a < 5);
7     a++;
8 }
```

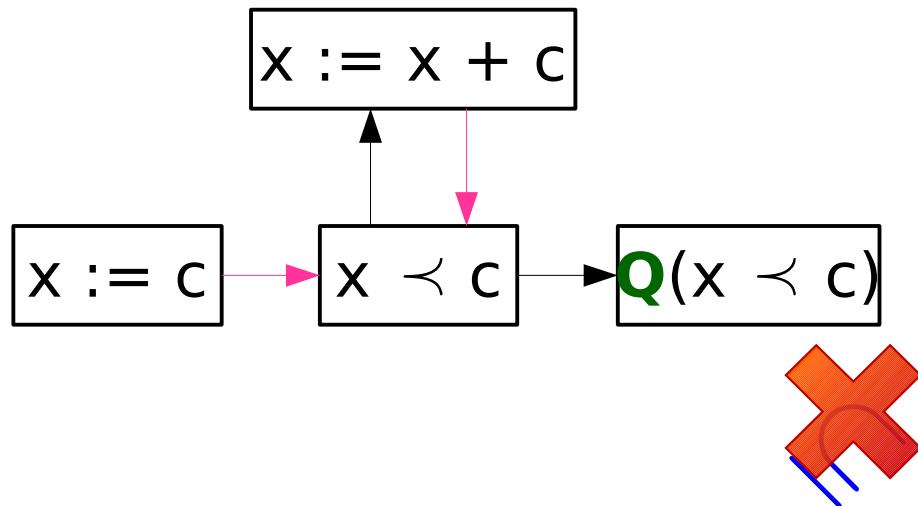
abstract data-flow graph:



Feature Check = Graph Inclusion Check

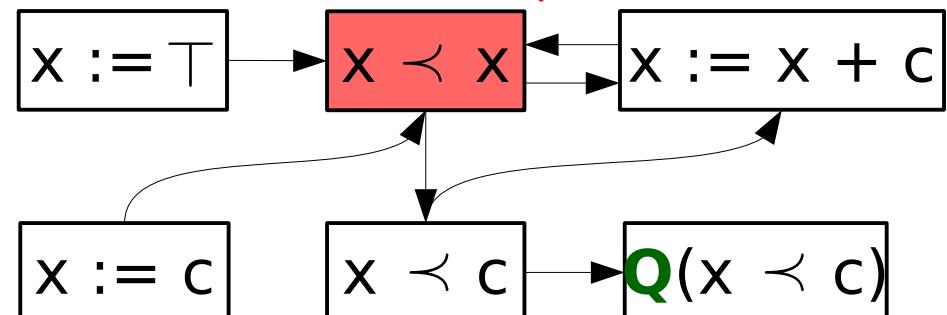
original program:

feature:



```
1  a = 0; b = 0;  
2  while (1) {  
3      b = unknown();  
4      if (a > b)  
5          if (a < 3)  
6              assert (a < 5);  
7      a++;  
8 }
```

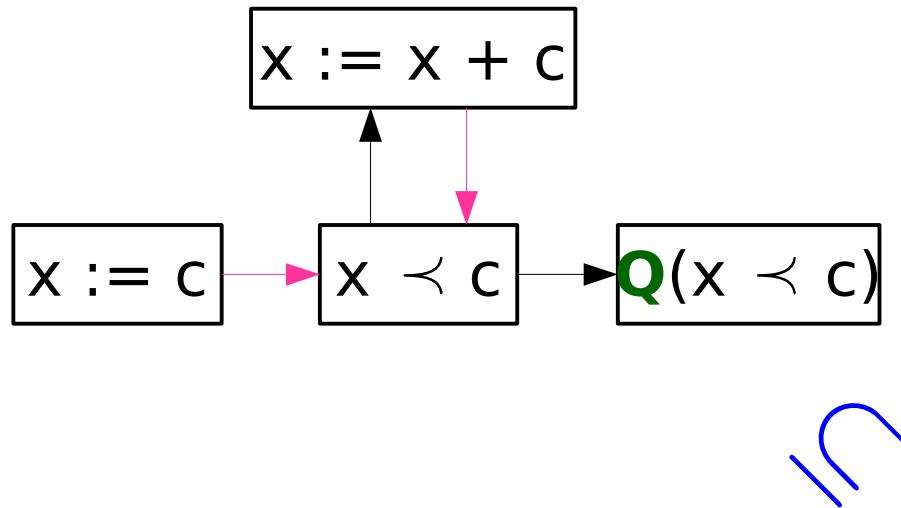
abstract data-flow graph:



Feature Check = Graph Inclusion Check

original program:

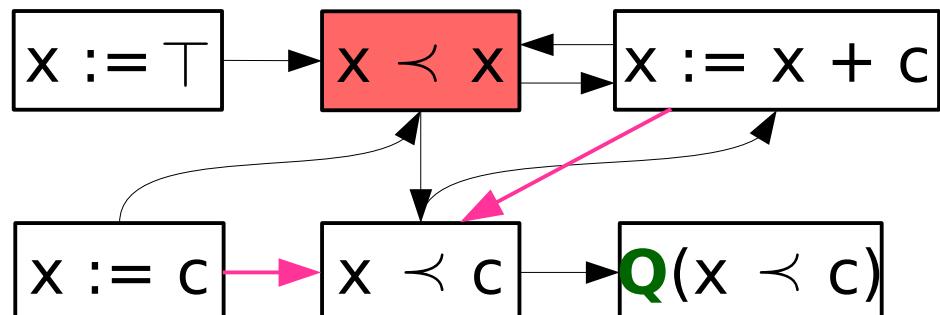
feature:



```
1  a = 0; b = 0;  
2  while (1) {  
3      b = unknown();  
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8 }
```

transitive closure

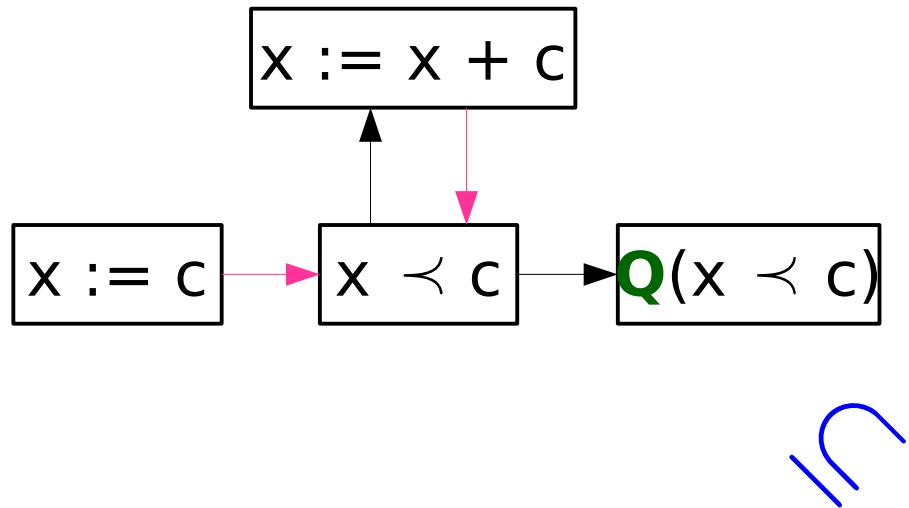
abstract data-flow graph:



Feature Check = Graph Inclusion Check

original program:

feature:



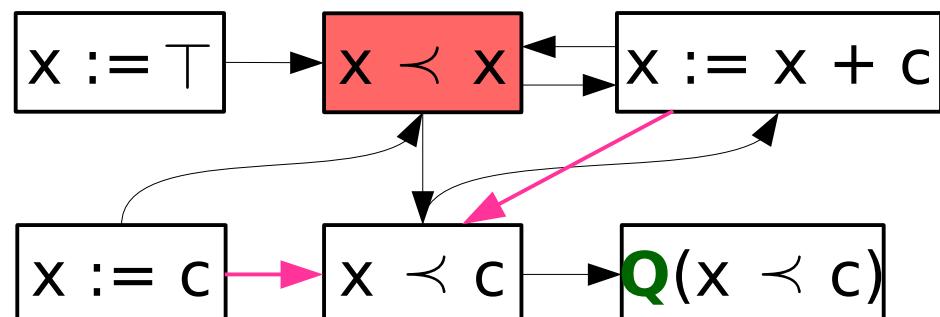
```
1  a = 0; b = 0;
2  while (1) {
3      b = unknown();
4      if (a > b)
5          if (a < 3)
6              assert (a < 5);
7      a++;
8 }
```

transitive closure

abstract data-flow graph:

Removing noise:

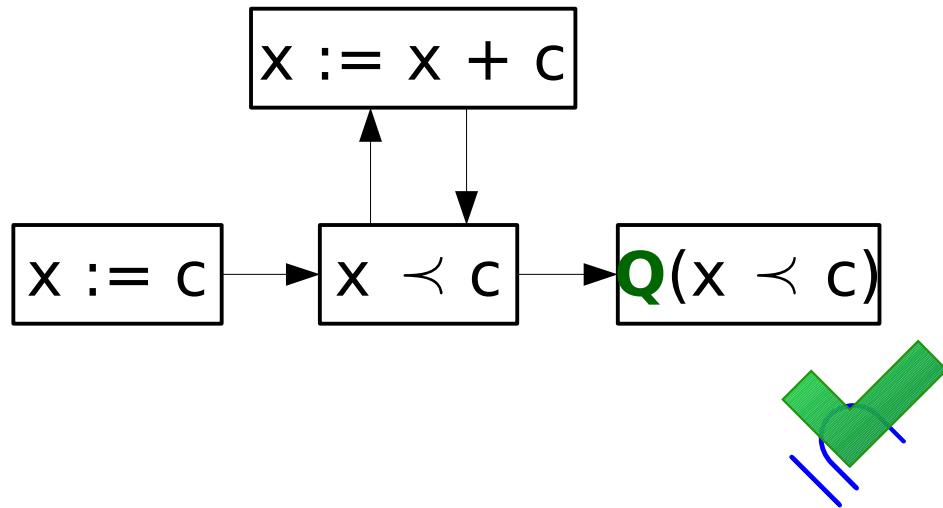
- ✓ reducer (offline, feature)
- ✓ transitive closure (online, new pgm)



Feature Check = Graph Inclusion Check

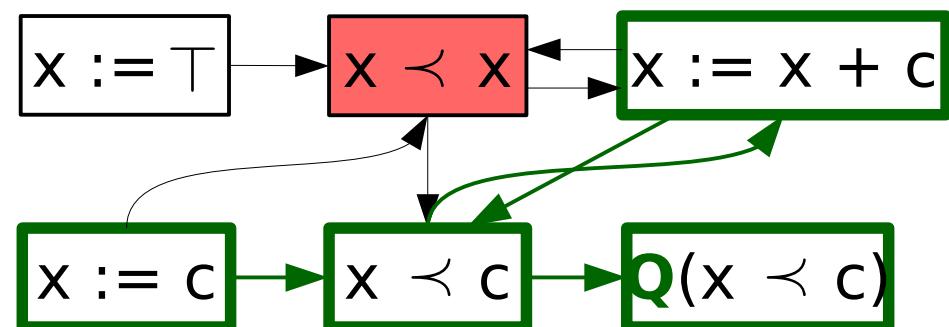
original program:

feature:



```
1  a = 0; b = 0;  
2  while (1) {  
3      b = unknown();  
4      if (a > b)  
5          if (a < 3)  
6              assert (a < 5);  
7      a++;  
8 }
```

abstract data-flow graph:



Evaluation

- Static analyzer:  (<https://github.com/ropas/sparrow>)
- Reducer: C-Reduce [PLDI'12] (<https://embed.cs.utah.edu/creduce>)
- Three instance analyses for C
 - Partially flow-sensitive interval analysis
 - Partially flow-sensitive pointer analysis
 - Partial Octagon analysis
- 60 benchmark programs from Linux and GNU packages

Results: Effectiveness (Classifier)

- Partially flow-sensitive interval analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost (sec)			Quality		(Oh et al. 2015)	
	Precision	Recall	FIr (a)	FSI (b)	Ours (c)	FIr (d)	FSI	Ours (e)	Prove	Cost	Prove	Cost
1	71.5 %	78.9 %	6,537	7,126	7,019	26.7	569.0	52.0	81.8 %	1.9x	56.6 %	2.0x
2	60.9 %	75.1 %	4,127	4,544	4,487	58.3	654.2	79.9	86.3 %	1.4x	49.2 %	2.4x
3	78.3 %	74.0 %	6,701	7,532	7,337	50.9	6,175.2	167.5	76.5 %	3.3x	51.1 %	3.4x
4	73.0 %	76.2 %	4,399	4,956	4,859	36.9	385.1	44.9	82.6 %	1.2x	54.8 %	1.2x
5	83.2 %	75.4 %	5,676	6,277	6,140	31.7	1,740.3	61.6	77.2 %	1.9x	65.6 %	1.8x
TOTAL	74.5 %	75.8 %	27,440	30,435	29,842	204.9	9,523.9	406.1	80.2 %	2.0x	55.1 %	2.3x

- Partially flow-sensitive pointer analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost (sec)			Quality	
	Precision	Recall	FIP	FSP	Ours	FIP	FSP	Ours	Prove	Cost
1	79.2 %	76.8 %	4,399	6,346	6,032	48.3	3,705.0	150.0	83.9 %	3.1x
2	78.3 %	77.2 %	7,029	8,650	8,436	48.9	651.4	74.0	86.8 %	1.5x
3	74.6 %	75.0 %	8,781	10,352	10,000	41.5	707.0	59.4	77.6 %	1.4x
4	73.9 %	76.0 %	10,559	12,914	12,326	51.1	4,107.0	164.3	75.0 %	3.2x
5	78.0 %	82.5 %	4,205	5,705	5,482	23.0	847.2	56.7	85.1 %	2.5x
TOTAL	76.7 %	77.4 %	34,973	43,967	42,276	212.9	10,017.8	504.6	81.2 %	2.4x

- Partial Octagon analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost (sec)			Quality		(Heo et al. 2016)	
	Precision	Recall	FSI	IMPCT	Ours	FSI	IMPCT	Ours	Prove	Cost	Prove	Cost
1	74.8 %	81.3 %	3,678	3,806	3,789	140.7	389.8	230.5	86.7 %	1.6 x	100.0 %	3.0 x
2	84.1 %	82.6 %	5,845	6,004	5,977	613.5	18,022.9	782.9	83.0 %	1.3 x	94.3 %	1.8 x
3	82.8 %	73.0 %	1,926	2,079	2,036	315.2	2,396.9	416.0	71.9 %	1.3 x	92.2 %	1.1 x
4	77.6 %	85.2 %	2,221	2,335	2,313	72.7	495.1	119.9	80.7 %	1.6 x	100.0 %	2.0 x
5	71.6 %	78.4 %	2,886	2,962	2,944	148.9	557.2	209.7	76.3 %	1.4 x	96.1 %	2.3 x
TOTAL	79.0 %	79.9 %	16,556	17,186	17,067	1,291.0	21,861.9	1,759.0	81.1 %	1.4 x	96.2 %	1.8 x

Results: Effectiveness (Analysis)

- Partially flow-sensitive interval analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost (sec)			Quality		(Oh et al. 2015)	
	Precision	Recall	FIr (a)	FSI (b)	Ours (c)	FIr (d)	FSI	Ours (e)	Prove	Cost	Prove	Cost
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- Partially flow-sensitive pointer analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost (sec)			Quality	
	Precision	Recall	FIP	FSP	Ours	FIP	FSP	Ours	Prove	Cost
1	79.2 %	76.8 %	4,399	6,346	6,032	48.3	3,705.0	150.0	83.9 %	3.1x
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- Partial Octagon analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost (sec)			Quality		(Heo et al. 2016)	
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TOTAL	79.0 %	79.9 %	16,556	17,186	17,067	1,291.0	21,861.9	1,759.0	81.1 %	1.4 x	96.2 %	1.8 x

Results: Comparison

- Partially flow-sensitive interval analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost (sec)			Quality		(Oh et al. 2015)	
	Precision	Recall	FIr (a)	FSI (b)	Ours (c)	FIr (d)	FSI	Ours (e)	Prove	Cost	Prove	Cost
1	71.5 %	78.9 %	6,537	7,126	7,019	26.7	569.0	52.0	81.8 %	1.9x	56.6 %	2.0x
2	60.9 %	75.1 %	4,127	4,544	4,487	58.3	654.2	79.9	86.3 %	1.4x	49.2 %	2.4x
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5	83.2 %	75.4 %	5,676	6,277	6,140	31.7	1,740.3	61.6	77.2 %	1.9x	65.6 %	1.8x
TOTAL	74.5 %	75.8 %	27,440	30,435	29,842	204.9	9,523.9	406.1	80.2 %	2.0x	55.1 %	2.3x

- Partial Octagon analysis

Trial	Query Prediction		#Proved Queries			Analysis Cost			Quality		(Hercules vs. manual)	
	Precision	Recall	FSI	IMPCT	Ours	FSI	IMPCT	Ours	Prove	Cost	Prove	Cost
1	74.8 %	81.3 %	3,678	3,806	3,789	140.7	389.8	100.0 %	1.6 x	100.0 %	1.6 x	1.6 x
2	84.1 %	82.6 %	5,845	6,004	5,977	613.5	18,022.9	782.9	93.0 %	1.3 x	94.3 %	1.3 x
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TOTAL	79.0 %	79.9 %	16,556	17,186	17,067	1,291.0	21,861.9	1,759.0	81.1 %	1.4 x	96.2 %	1.8 x

- Consistently perform well on a wide range of programs. (\leftrightarrow wide variation)
- No clear conclusion (different approaches and learning algorithms)

Results: Generated Features (Top 2)

- Partially flow-sensitive interval analysis

feature program 1:

```
int buf [10];
for (i = 0; i < 7; i++) {
    buf[i] = 0; // Query
}
```

feature program 2:

```
k = 255; p = malloc (k);
while (k > 0) {
    *(p + k) = 0; // Query
    k--;
}
```

- Access to a consecutive memory region in a loop
- Bounded indice by a constant

Results: Generated Features (Top 2)

- Partially flow-sensitive pointer analysis

feature program 1:

```
int j = 16; q = malloc(j)
if (q == 0)
    return;
else *q = 0; // Query
```

feature program 2:

```
r = malloc(v);
r = &a;
*r = 0; // Query
```

- Null-check before buffer access
- Strong update by the address of another variable

Results: Generated Features (Top 2)

- Partial Octagon analysis

feature program 1:

```
size = POS_NUM;  
arr = malloc(size);  
arr[size-1] = 0; // Query
```

feature program 2:

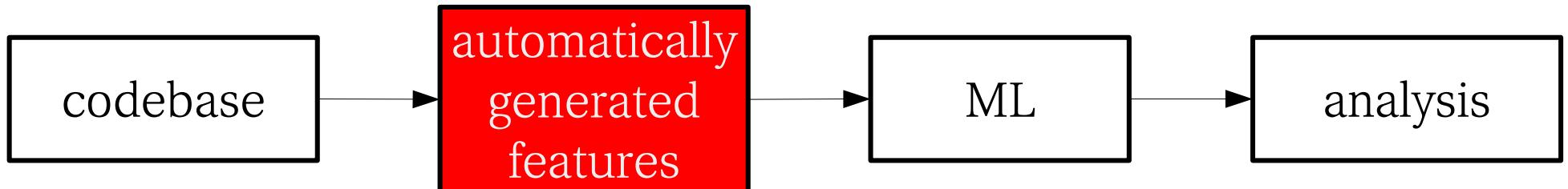
```
idx = POS_NUM;  
buf = malloc(idx);  
for (n = 0; n < idx; n++) {  
    buf[n] = 0; // Query  
}
```

- Array of a **positive size**
 - e.g., when $\text{POS_NUM} = [1, +\infty]$ in the flow-sensitive interval analysis
 - Index related to the size **in a simple linear way**

Caveats: Expressiveness of Features

- Our feature representation is expressive enough, but not perfect, e.g.,
 - ✓ “x and y results in finite intervals after analysis.” 
 - ✓ “ 2^k type of integers are important constants.” 

Summary



- “Features” in data-driven static analysis
 - By reducing programs
 - As generalized graphs (\leftrightarrow program text)