

# Improving Static Analysis Precision by Minimal Program Refinement

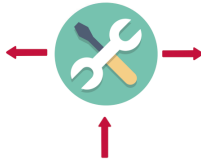
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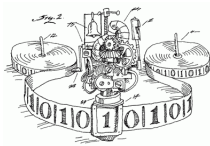
(Work in Progress)

## Context



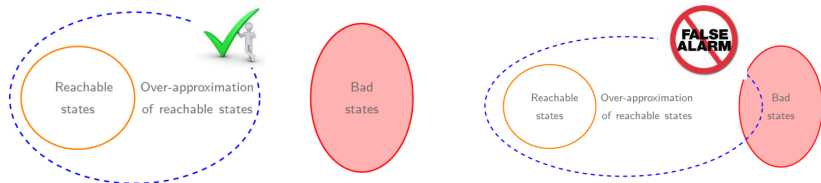
# (Sound) Static Analysis

```
int i=i0;  
while (i<100)  
  j=0;  
  while(j<i)  
    j=j+1  
    i=i+1;  
assert(i==100);
```



```
{i0>=0}  
int i=i0;  
{i=i0, i0>=0}  
while (i<100)  
  {i=i0, i<100, i0>=0}  
  j=0;  
  {i=i0, i<100, j=0, i0>=0}  
  while(j<i)  
    {i=i0, i<100, j<i, i0>=0}  
    j=j+1  
    {i=i0, i<100, j<=i, i0>=0}  
    {i=i0, i<100, j=i, i0>=0}  
    i=i+1;  
    {i+1=i0, i<=100, j=i+1, i0>=0}  
  
  {i=100, i0>=0}  
  assert(i==100);  
  {i=100, i0>=0}
```

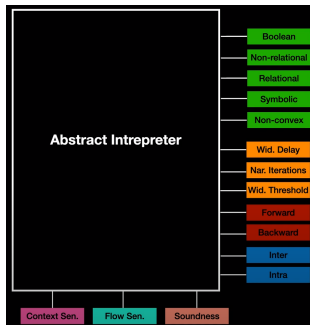
# (Sound) Static Analysis



Imprecision is a common phenomenon which raises false alarms

**Goal:** To improve precision of a static analysis

# Problem: Improving Precision is tricky



- ▶ Many parameters to tune
- ▶ Creating new domains requires lots of efforts and expertise

Hence the need for **automatic** precision refinement

**Goal:** To design an automatic refinement framework for static analysis which is:

1. **Generic:** Independent from the abstract domain
2. **Principled:** General strategy to control disjunctions

# Automatic Refinement

**Goal:** To design an automatic refinement framework for static analysis which is:

1. **Generic:** Independent from the abstract domain
2. **Principled:** General strategy to control disjunctions

		Generic	Principled
Model-Checking		✗ Specialized E.g. CEGAR	✓
Static Analysis	Trace Partitioning	✓	✗ Heuristics
	Others	✗ Specialized X Rival. in Shape analysis, etc	✓

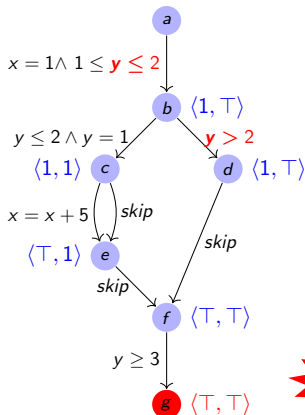
## Contributions

- ▶ A new automatic refinement framework
  - Refinements are CFG-based
  - Independent of any abstract domain ( $\sqcup, \sqcap, \text{post}, \llbracket \cdot \rrbracket^*$ )
- ▶ Identify desired properties of the framework
  - Correctness, Completeness, Minimality
  - Interface contracts
- ▶
  - Optimal instance with the above all 3 desired properties
  - Optimizations
    - ✓ E.g state-space Exploration, Incremental computation
- ▶ Handle non-linear reasoning (widening)



# Program Refinement

Constant propagation analysis with abstract states  $\{\langle x, y \rangle\}$

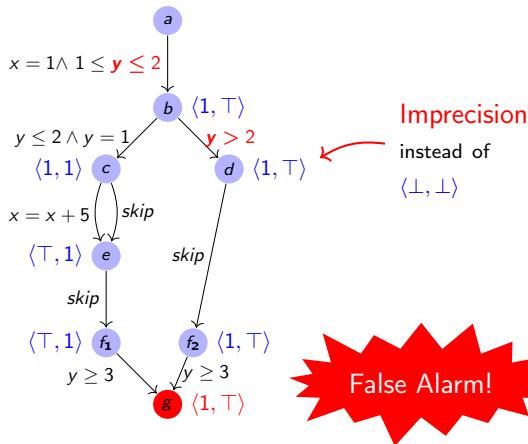


False Alarm!

Imprecision at  $g$ . Can we remove it?

# Program Refinement

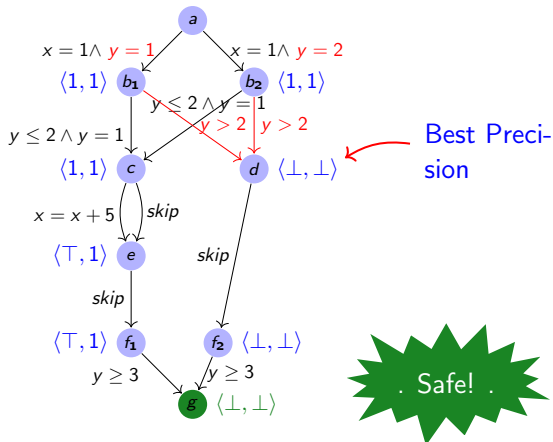
Constant propagation analysis with abstract states  $\{\langle x, y \rangle\}$



Split at  $f$  reduces imprecision at  $g$ . Can we reduce more?

# Program Refinement

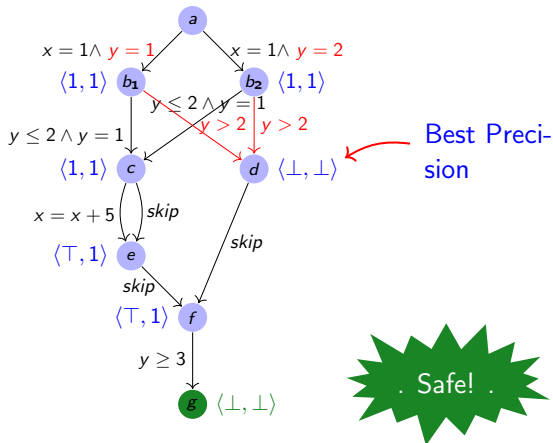
Constant propagation analysis with abstract states  $\{\langle x, y \rangle\}$



Can some of the splits be merged?

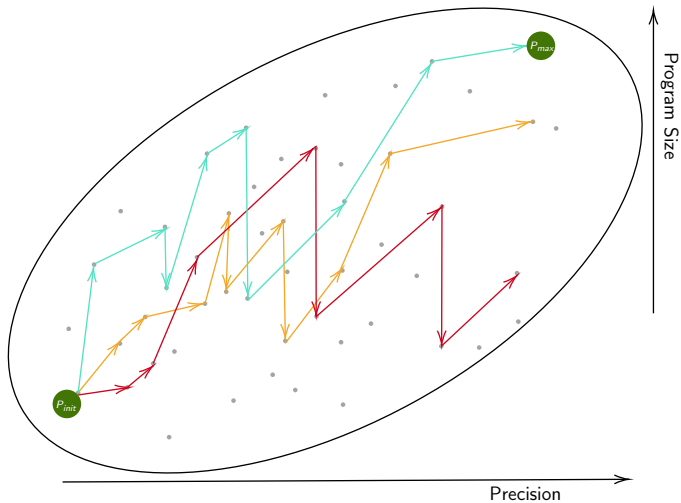
# Program Refinement

Constant propagation analysis with abstract states  $\{\langle x, y \rangle\}$



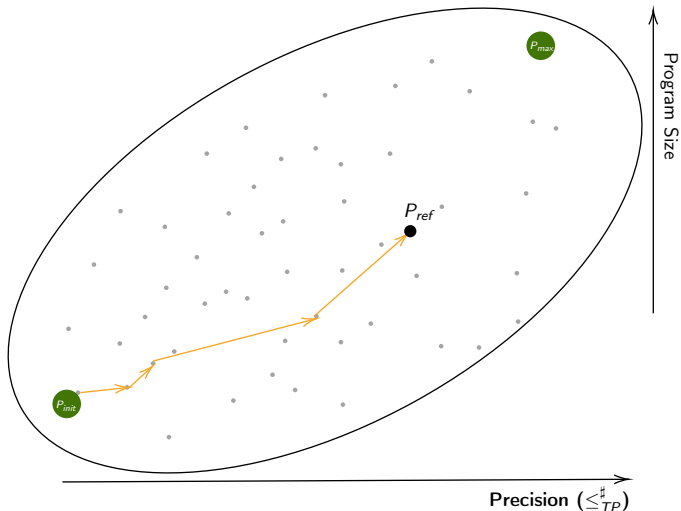
Yes, and there is no loss of precision

# Generic Refinement Techniques



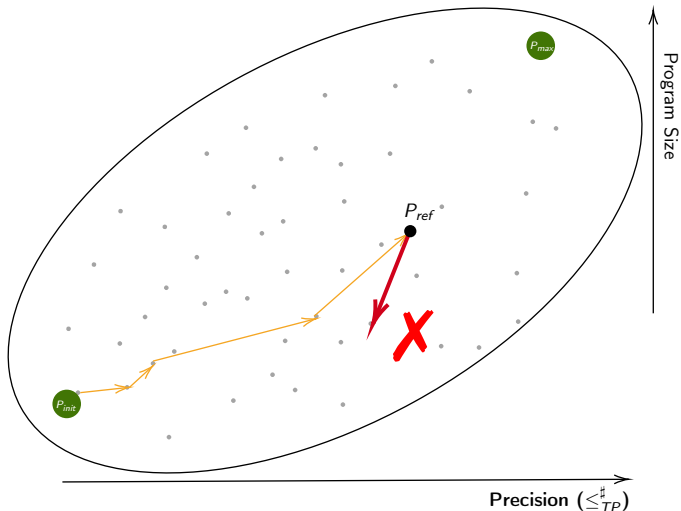
# Problem of Standard Trace Partitioning Framework 1

**Problem.** Decreasing size leads to loss of precision.



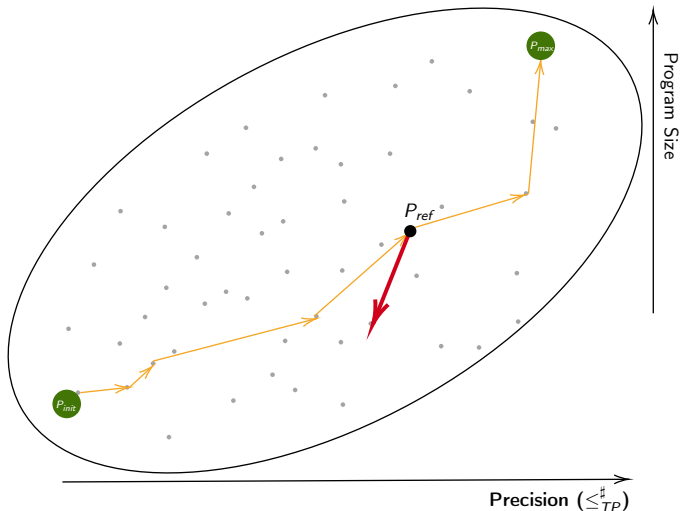
# Problem of Standard Trace Partitioning Framework 1

**Problem.** Decreasing size leads to loss of precision.



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**Problem.** Decreasing size leads to loss of precision.





# Problem of Standard Trace Partitioning Framework 2

**Problem.** Decreasing size leads to loss of precision.

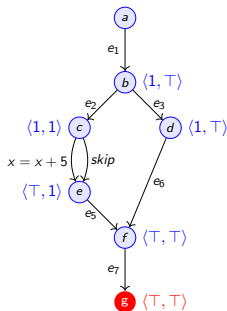


Figure:  $P_1$

*split*  
→

←  
*merge*

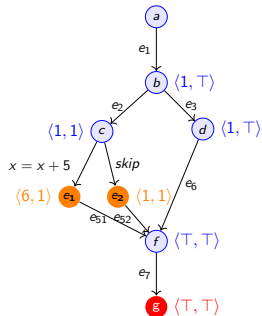


Figure:  $P_2$

# Problem of Standard Trace Partitioning Framework 2

**Problem.** Decreasing size leads to loss of precision.

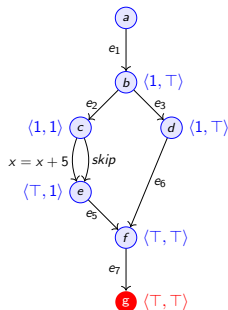


Figure:  $P_1$

split  
→

←  
merge

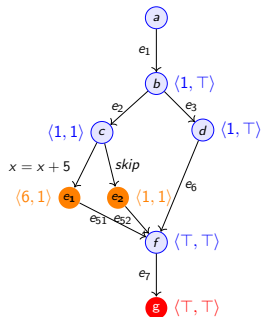
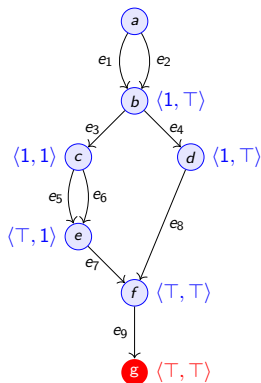


Figure:  $P_2$

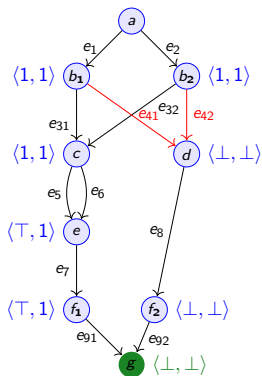
**Solution.** Choose *locations of interest*  $\mathcal{L}$ .

✓ E.g. For  $\mathcal{L} = \{g\}$ , merge does not lose precision at  $g$

# Refinements as $n$ -tuples

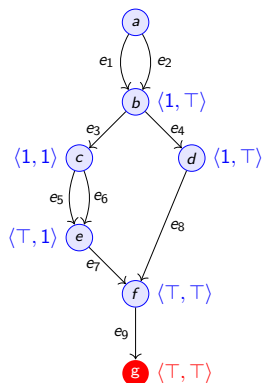


$$t = [(e_7, e_8, 1), (e_1, e_2, 1)]$$

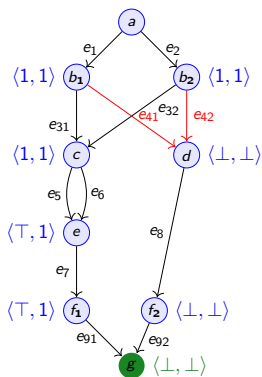


- $t \in \mathbb{N}^n$
- Partial Orderings on  $n$ -tuples:
  1. Component-wise ordering  $\leq$
  2. Precision ordering  $\leq_{\mathcal{L}}^{\#}$

# Refinements as $n$ -tuples

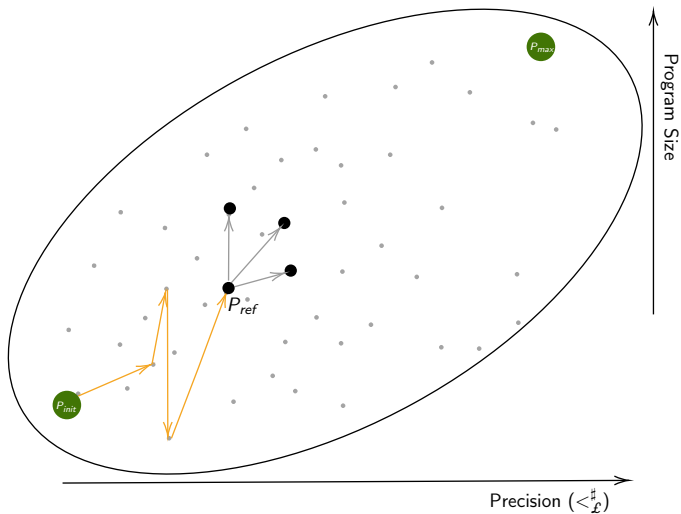


$$t = (1, 1)$$

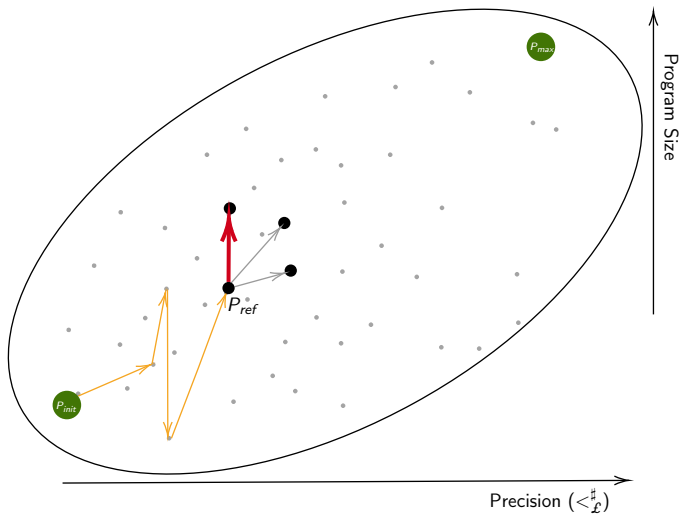


- $t \in \mathbb{N}^n$
- Partial Orderings on  $n$ -tuples:
  1. Component-wise ordering  $\leq$
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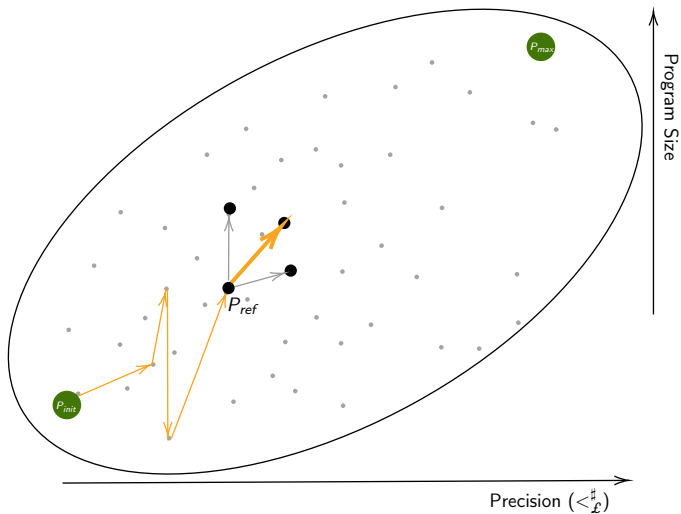
# Generic Refinement Framework: **next** method



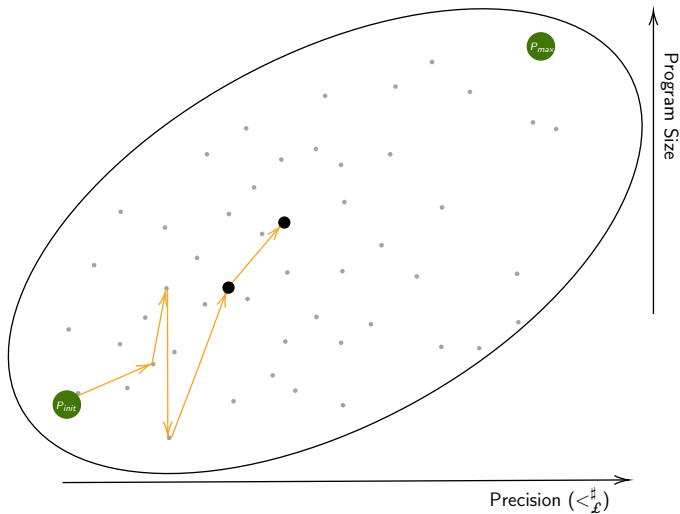
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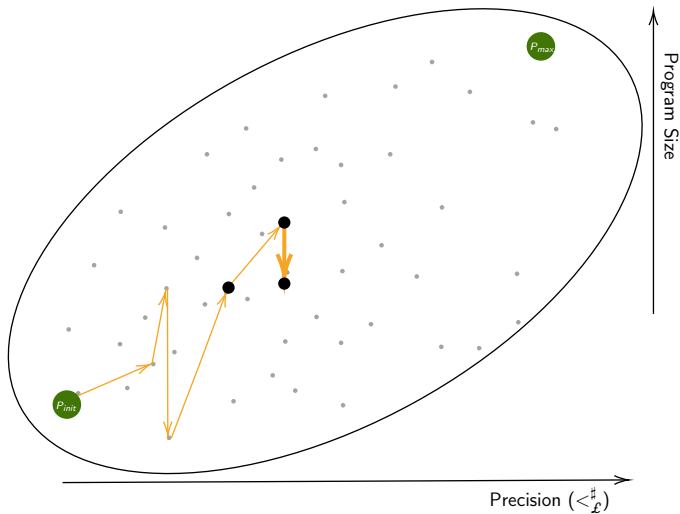


# Generic Refinement Framework: **collapseSplits** method

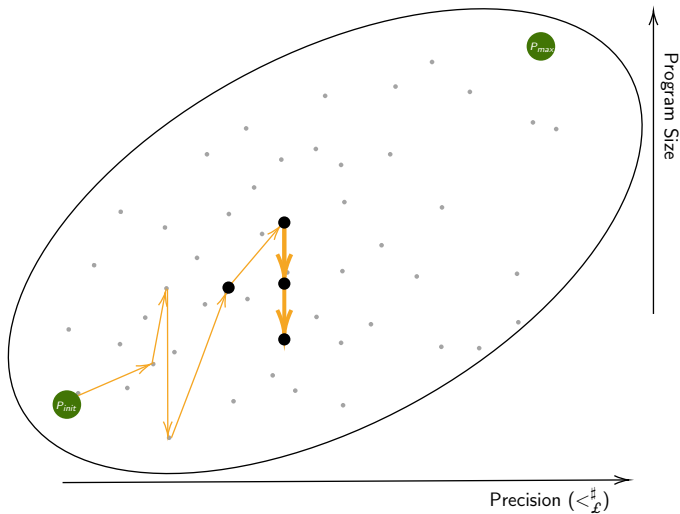




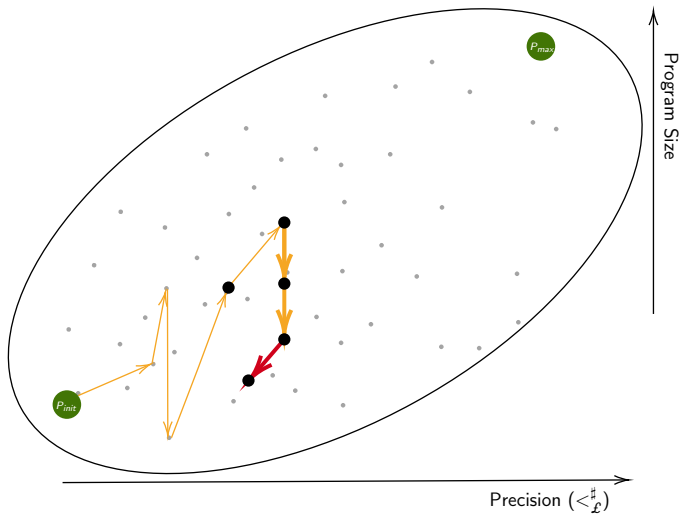
# Generic Refinement Framework: **collapseSplits** method



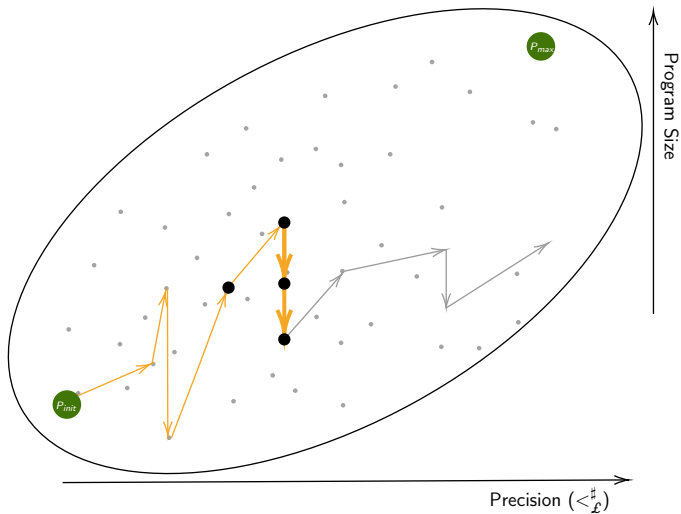
# Generic Refinement Framework: **collapseSplits** method



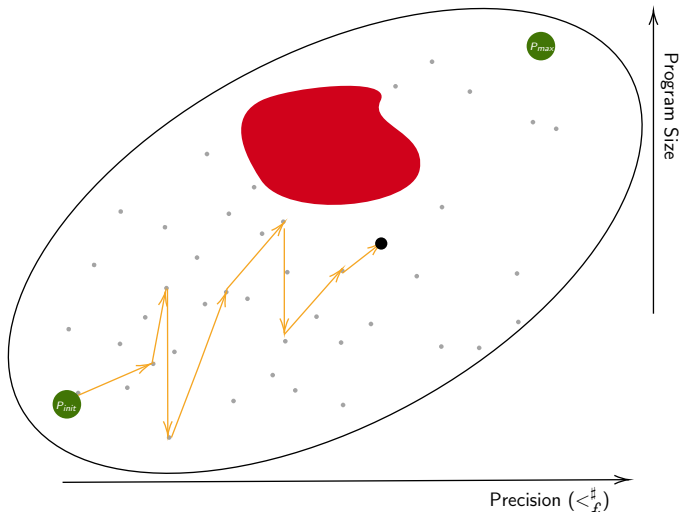
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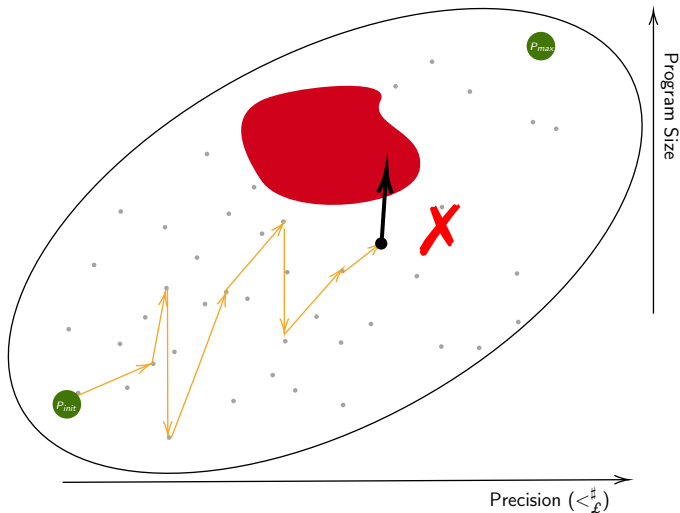


# Generic Refinement Framework: Optimizing Exploration



next prunes search space with region set **AVOID**

# Generic Refinement Framework: Optimizing Exploration



next prunes search space with region set **AVOID**

# Generic Refinement Framework

**Input** : Program  $P$ ,  $D^\#$ , max dist  $k$ , loc  $\mathcal{L}$

**Output**: Tuple with maximum precision

```
1   $t_{prev} = t_{best} = (0, \dots, 0)$ ,  $AVOID = \emptyset$ ;  
2  while (true) do  
3       $t_{cur} := \text{next}(t_{prev}, t_{best}, AVOID)$  ;  
4      if  $t_{cur} == \text{null}$  then break; ;  
5      else  
6          if  $t_{cur} <_{\mathcal{L}}^\# t_{best}$  then  
7               $t_{best} =$   
                 $\text{collapseSplit}(t_{best}, t_{cur})$ ;  
8          else  
9               $AVOID = AVOID \cup$   
                 $\text{newPruneSet}(t_{cur}, t_{best}, i, AVOID)$ ;  
10          $t_{prev} = t_{cur}$ ;  
11 return  $t_{best}$ ;
```

## Variables.

1. Previous tuple  $t_{prev}$
2. Current tuple  $t_{cur}$
3. Best tuple  $t_{best}$
4. Pruning set  $AVOID$

## Definition (Correctness, global improvement)

Termination + Global Improvement

## Definition (Completeness)

Finds a maximally precise refinement

## Definition (Minimality)



# When do we have Algorithm properties?

	Correctness	Completeness	Minimality
<code>next</code>	<b>Monotonicity</b> (Global Imp)	-	-
<code>collapseSplits</code>	Merge-Lossless (Termination)	Merge-Lossless	Merge-Minimal
<code>newPruneSet</code>	-	Non-Improving	-

# Comparison

	Correctness	Completeness	Minimality
Naive GSR	✓	✗	✗
Trace partitioning	✓	✗	✗
Optimal GSR	✓	✓	✓
Intermediate GSR	✓	✓	✗

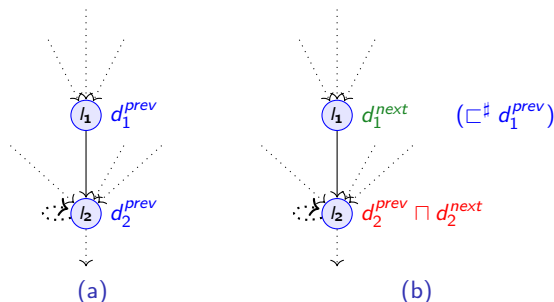
# Problem with Monotonicity Assumption

Widening is **non-monotone**:

$$[1, 40] \nabla [1, 40] = [1, 40] \quad \xRightarrow{\text{splits}} \quad [1, 1] \nabla [1, 40] = [1, +\infty]$$

# Monotonicity Forcing Trick

**Key Idea.** Take previous refinement fixpoint map and force monotonicity.



**Figure:** a) Previous Refinement; b) Forcing monotonicity in the new refinement

Both  $d_2^{prev}$  and  $d_2^{next}$  are over-approximation of reachable states.  
*So, take a meet  $\sqcap$  of them*

Within static analysis we provide a new refinement framework that is automatic, generic and principled.

Thanks for your attention!