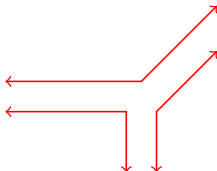


# Using Dynamic Analysis to Generate Disjunctive Invariants

ThanhVu (Vu) Nguyen\*,  
Deepak Kapur\*, Westley Weimer<sup>†</sup>, Stephanie Forrest\*

\*University of New Mexico, <sup>†</sup>University of Virginia

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# Program Invariants

- *“Invariants are asserted properties, such as relations among variables, at certain locations in a program”*
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  - Pre/Post conditions
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- Invariants can generated using static or dynamic analysis
  - Static analysis examines program source code
  - Dynamic learns from program execution traces

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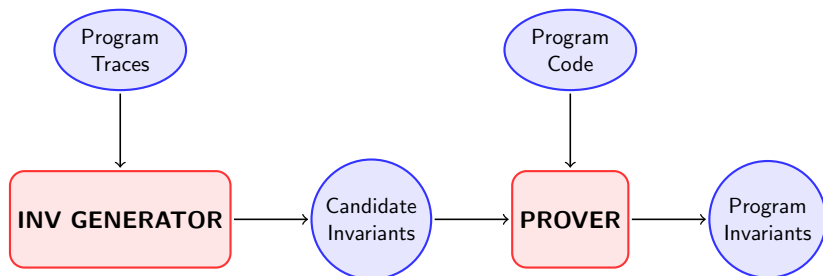
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  - Represent semantics of branching  
The invariant after `if (p) {a=1;} else {a=2;} is`  
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  - Existing approaches focus mostly on *conjunctive* relations
- Existing approaches have *trade-offs* among soundness, efficiency, and expressive power
  - Static analyzers, e.g., Interproc, Astrée, support conjunctive relations
  - Dynamic techniques, e.g., Daikon, also have limited support for disjunctive invariants and can produce spurious results



# Hybrid Invariant Generation



- **DIG (Dynamic Invariant Generator)**
  - Find invariants directly from program traces
  - Build **nonconvex polyhedra** over trace points and extract facets representing disjunctive relations
- **KIP (K-Inductive Prover)**
  - Verify candidate invariants statically from program code
  - Based on **k-induction** and SMT solving

# Geometric Invariant Inference

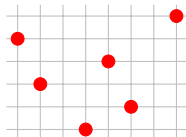
- Treat trace values as points in multi-dimensional space
- Build a **convex hull** (polyhedron) over the points
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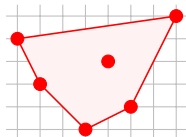
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x	y
-2	1
-1	-1
1	-3
2	0
3	-2
5	2

program traces



trace pts  
in 2D



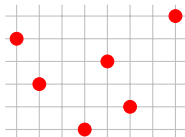
polygon  
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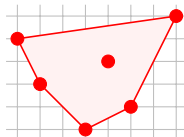
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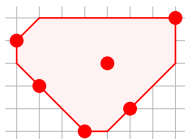


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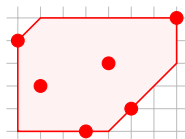


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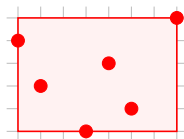
- Consider simpler shapes (decreasing precision, increasing efficiency)



octagon  
 $\pm x \pm y \geq c$



zone  
 $x - y \geq c$



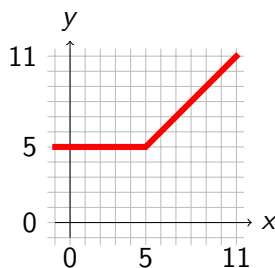
box  
 $\pm x, y \geq c$

# Outline

## Example

```
def ex(x):  
    y = 5  
    if x > y: x = y  
    while[L] x ≤ 10:  
        if x ≥ 5:  
            y = y + 1  
            x = x + 1  
  
    assert y ≡ 11
```

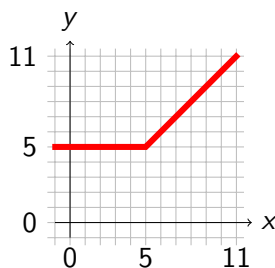
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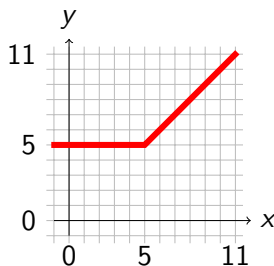


$$L : (x < 5 \wedge 5 = y) \vee (x \geq 5 \wedge x = y), 11 \geq x$$

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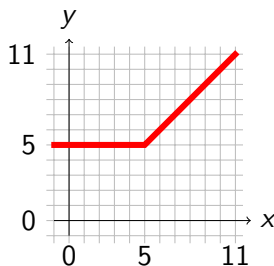
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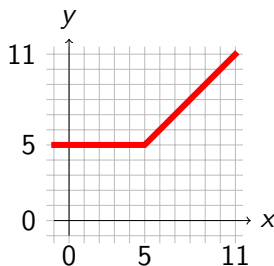
② if  $x \geq 5$  then  $x = y$

$$\longleftrightarrow \text{if } 0 > x - 5 \text{ then } 0 = y - 5 \text{ else } x - 5 = y - 5$$
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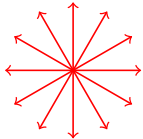
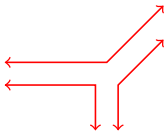
a linear relation .. in **max-plus algebra**

# Max-plus Algebra

# Max-plus Algebra

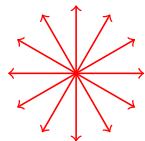
	Linear	Max-plus
Domain	$\mathbb{R}$	$\mathbb{R} \cup \{-\infty\}$
Addition	$+$	$\max$
Multiplication	$\times$	$+$
Zero elem	$0$	$-\infty$
Unit elem	$1$	$0$
Relation form	$c_0 + c_1 t_1 + \dots + c_n t_n \geq 0$	$\max(c_0, c_1 + t_1, \dots, c_n + t_n) \geq \max(d_0, d_1 + t_1, \dots, d_n + t_n)$

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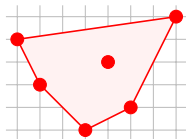
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Line shapes		

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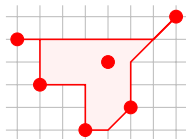
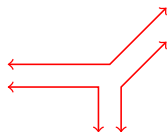
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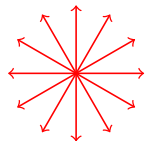


Convex hull

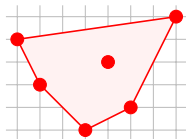


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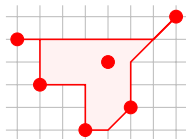
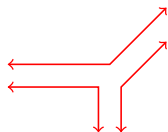
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Line shapes



Convex hull



Not convex in classical sense!

# Dynamically Inferring Max-plus Invariants

## Examples

$$z = \max(x, y) \quad \equiv \quad (x < y \wedge z = x) \quad \vee \quad (x \geq y \wedge z = y)$$



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DIG discovers disjunctive relations of the **max-plus** form

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## Method

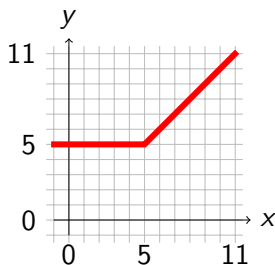
- Represent trace values as points
- Build a max-plus convex polyhedron
- Extract facets represented by max-plus relations

## Example

$x$	$y$
-1	5
$\vdots$	$\vdots$
5	5
6	6
$\vdots$	$\vdots$
11	11

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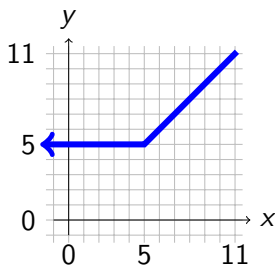
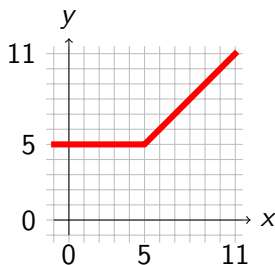
- DIG builds a max-plus polygon and finds candidate invs:

$$11 \geq x \geq -1, \quad 11 \geq y \geq 5, \quad 0 \geq x - y \geq -6$$

$$\max(0, x - 5) \geq y - 5 \equiv (x < 5 \wedge 5 \geq y) \vee (x \geq 5 \wedge y \leq x)$$

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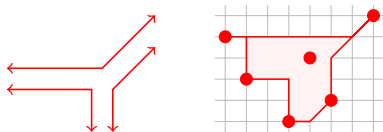
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- KIP removes the spurious invariants  $x \geq -1$  and  $x - y \geq -6$
- Remaining invariants are **true** and equivalent to

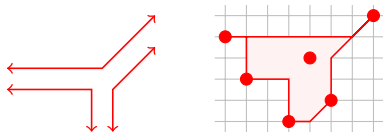
$$(x < 5 \wedge 5 = y) \vee (11 \geq x \geq 5 \wedge x = y)$$

## Weak Max-plus Invariants

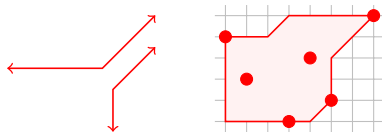


## General Max-plus

# Weak Max-plus Invariants



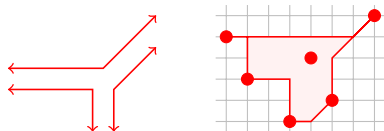
General Max-plus



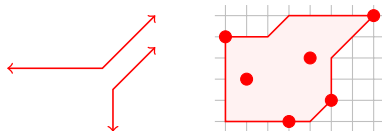
Weak Max-plus



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General Max-plus



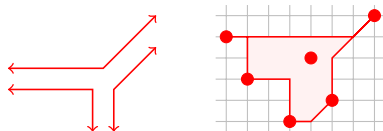
Weak Max-plus

- DIG introduces **weak** max-plus relations of the form

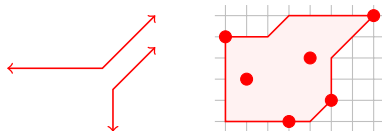
$$\begin{aligned} \max(c_0, c_1 + t_1, \dots, c_k + t_k) &\geq t_j + d, \\ t_j + d &\geq \max(c_0, c_1 + t_1, \dots, c_k + t_k), \end{aligned}$$

where  $d \in \mathbb{R}$  and  $t_j \in \{t_1, \dots, t_k\}$

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where  $d \in \mathbb{R}$  and  $t_j \in \{t_1, \dots, t_k\}$

- Restrict values of coefs  $c_i$  to  $\{0, -\infty\}$
- Fix the number  $k$  of variables  $t_i$ , e.g.,  $k = 2$
- Allow only one unknown param  $d$

## Algorithm for Finding Weak Max-plus Invs

Given 2D points  $\{(x_1, y_1), \dots, (x_n, y_n)\}$ , build a weak max-plus polygon, i.e., a conjunction of inequalities of the form

$$\begin{aligned} \max(c_0, c_1 + x, c_2 + y) &\geq x + d, & x + d &\geq \max(c_0, c_1 + x, c_k + y), \\ \max(c_0, c_1 + x, c_2 + y) &\geq y + d, & y + d &\geq \max(c_0, c_1 + x, c_k + y) \end{aligned}$$

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### ① Enumerate weak relations by instantiating $c_i$ over $\{0, -\infty\}$

- Each weak max-plus form yields at most 8 relations,  
e.g.,  $\max(c_0, c_1 + x, c_2 + y) \geq x + d$  produces

$$\begin{aligned} \max(0, x, y) &\geq x + d, & \max(0, x) &\geq x + d, \\ \max(0, y) &\geq x + d, & 0 &\geq x + d, \dots \end{aligned}$$

- Obtain at most 32 relations from 4 weak max-plus forms

## Algorithm for Finding Weak Max-plus Invs

Given 2D points  $\{(x_1, y_1), \dots, (x_n, y_n)\}$ , build a weak max-plus polygon, i.e., a conjunction of inequalities of the form

$$\begin{aligned} \max(c_0, c_1 + x, c_2 + y) &\geq x + d, & x + d &\geq \max(c_0, c_1 + x, c_k + y), \\ \max(c_0, c_1 + x, c_2 + y) &\geq y + d, & y + d &\geq \max(c_0, c_1 + x, c_k + y) \end{aligned}$$

① **Enumerate** weak relations by instantiating  $c_i$  over  $\{0, -\infty\}$

- Each weak max-plus form yields at most **8** relations,  
e.g.,  $\max(c_0, c_1 + x, c_2 + y) \geq x + d$  produces

$$\begin{aligned} \max(0, x, y) &\geq x + d, & \max(0, x) &\geq x + d, \\ \max(0, y) &\geq x + d, & 0 &\geq x + d, \dots \end{aligned}$$

- Obtain at most **32** relations from 4 weak max-plus forms

② **Solve** for  $d$  in the obtained relations using given points, e.g.,

$$\begin{aligned} \max(0, y) &\geq x + d &\rightarrow d &= \min(\max(0, y_i) - x_i) \\ x + d &\geq \max(0, y) &\rightarrow d &= \max(\max(0, y_i) - x_i) \end{aligned}$$

# Min-plus Invariants

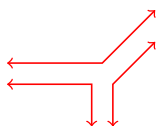
- DIG discovers disjunctive relations of the **min-plus** form

$$\min(c_0, c_1 + t_1, \dots, c_n + t_n) \geq \min(d_0, d_1 + t_1, \dots, d_n + t_n)$$

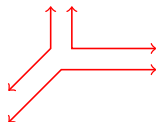
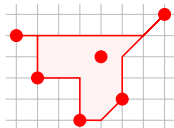
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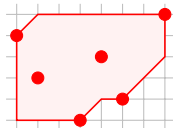
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Max-plus



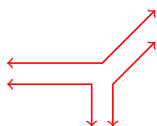
Min-plus



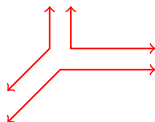
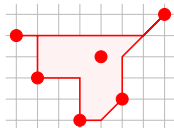
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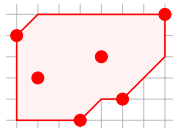
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Max-plus



Min-plus



- Also support **weak** min-plus relations of the form

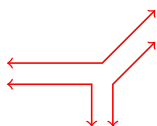
$$\begin{aligned} \min(c_0, c_1 + t_1, \dots, c_k + t_k) &\geq t_j + d_i, \\ t_j + d_i &\geq \min(c_0, c_1 + t_1, \dots, c_k + t_k) \end{aligned}$$



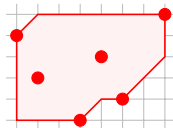
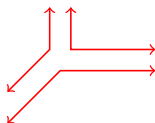
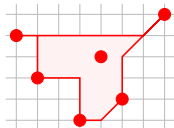
# Min-plus Invariants

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- Combine max and min-plus invariants for more expressive power, e.g., can capture iff behavior

# Algorithmic Analysis

	<b>Linear</b>	<b>Max/Min-plus</b>	<b>Weak Max/Min-plus</b>
Complexity	$O(p^{\frac{n}{2}})$	$O(pn^2(p+n)^n)$	$O(p2^k)$

$p$  = # of trace pts,  $n$  = # of variables,  $k$  = a predefined constant, e.g.,  $k = 2$

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## Underapproximation Property

- *Theorem*: If the form of the *true* invariant  $f$  is supported, then DIG generates only the candidate invariant  $f'$  such that  $f' \Rightarrow f$
- *Application*: useful for debugging; if  $f'$  *strictly* overapproximates  $f$ , then there is a trace representing a counterexample violating  $f$

# Outline

# Proving Program Invariants Using k-Induction

- Represent a program execution as a **state transition system**

$$M = (I, T),$$

with the initial state  $I$  and the transition relation  $T$

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- Use **k-induction** to prove that  $p$  is an invariant of  $M$

$$\begin{aligned} I \wedge T_1 \wedge \cdots \wedge T_k &\Rightarrow p_0 \wedge \cdots \wedge p_k \\ p_n \wedge T_{n+1} \wedge \cdots \wedge p_{n+k} \wedge T_{n+k+1} &\Rightarrow p_{n+k+1} \end{aligned}$$

- More powerful than standard ( $k = 0$ ) induction
- Help prove properties cannot be proved using standard induction

## Example

```
def sqrt(x):  
    assert(x ≥ 0);  
    a = 0; s = 1; t = 1  
    while[L] s ≤ x:  
        a += 1; t += 2; s += t  
    return a
```

### DIG

find candidate invariants at L

---

$$4s = t^2 + 2t + 1$$

$$t = 2a + 1$$

$$s = (a + 1)^2$$

$$s \geq t$$

$$9989 \geq x$$

⋮



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distinguish **true** and **spurious** invs

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spurious

⋮

# Features of KIP

- Iterative k-induction
- Employ the Z3 SMT solver
- Learn lemmas
- Eliminate redundancy
- Parallelism

# Outline

# Evaluation

## Benchmarks

- Disjunctive testsuite: 14 programs require disjunctive invariants
- Nonlinear test suite: 27 programs require nonlinear invariants

## Setup

- Implemented in SAGE/Python (with Z3 backend solver)
- Test machine: 64-core 2.6GHZ CPU, 128GB RAM, Linux OS
- Traces obtained at loop entrances and program exits

# Evaluation

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- Disjunctive testsuite: 14 programs require disjunctive invariants
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## Setup

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- Test machine: 64-core 2.6GHZ CPU, 128GB RAM, Linux OS
- Traces obtained at loop entrances and program exits

## Results

- All generated equalities are valid and most inequalities are spurious (and removed by KIP)
- Identified invariants are sufficiently strong to explain program behavior
- Current dynamic analysis cannot find any of these invariants



## Results for Disjunctive Invariants

Prog	Loc	Var
ex	1	2
strncpy	1	3
oddeven3	1	6
oddeven4	1	8
oddeven5	1	10
bubble3	1	6
bubble4	1	8
bubble5	1	10
partd3	4	5
partd4	5	6
partd5	6	7
parti3	4	5
parti4	5	6
parti5	6	7
<b>total</b>		

## Results for Disjunctive Invariants

Prog	Loc	Var	Gen	T <sub>Gen</sub> (secs)
ex	1	2	15	0.2
strncpy	1	3	69	1.1
oddeven3	1	6	286	3.7
oddeven4	1	8	867	12.7
oddeven5	1	10	2334	56.8
bubble3	1	6	249	4.1
bubble4	1	8	832	11.7
bubble5	1	10	2198	53.9
partd3	4	5	479	10.5
partd4	5	6	1217	23.3
partd5	6	7	2943	53.3
parti3	4	5	464	10.3
parti4	5	6	1148	22.4
parti5	6	7	2954	53.6
<b>total</b>			16055	317.6

## Results for Disjunctive Invariants

Prog	Loc	Var	Gen	T <sub>Gen</sub> (secs)	Val	T <sub>Val</sub> (secs)
ex	1	2	15	0.2	4	1.5
strncpy	1	3	69	1.1	4	7.7
oddeven3	1	6	286	3.7	8	16.0
oddeven4	1	8	867	12.7	22	46.0
oddeven5	1	10	2334	56.8	52	1319.4
bubble3	1	6	249	4.1	8	4.9
bubble4	1	8	832	11.7	22	47.6
bubble5	1	10	2198	53.9	52	938.2
partd3	4	5	479	10.5	10	50.8
partd4	5	6	1217	23.3	15	181.1
partd5	6	7	2943	53.3	21	418.1
parti3	4	5	464	10.3	10	45.5
parti4	5	6	1148	22.4	15	165.1
parti5	6	7	2954	53.6	21	405.6
<b>total</b>			16055	317.6	264	3647.5

## Results for Disjunctive Invariants

Prog	Loc	Var	Gen	T <sub>Gen</sub> (secs)	Val	T <sub>Val</sub> (secs)	Strength
ex	1	2	15	0.2	4	1.5	✓
strncpy	1	3	69	1.1	4	7.7	✓
oddeven3	1	6	286	3.7	8	16.0	✓
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bubble5	1	10	2198	53.9	52	938.2	✓
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<b>total</b>			16055	317.6	264	3647.5	<b>14/14</b>

# Results for Complex Invariants

Prog	Loc	Var	Gen	T <sub>Gen</sub> (secs)	Val	T <sub>Val</sub> (secs)	kl	Strength
cohendv	2	6	152	26.2	7	8.2	14	✓
divbin	2	5	96	37.7	8	8.7	15	—
manna	1	5	49	19.2	3	5.6	2	✓
hard	2	6	107	14.2	11	9.2	4	—
sqrt1	1	4	27	25.3	3	4.3	1	✓
dijkstra	2	5	61	30.7	8	10.9	6	—
freire1	1	3	25	22.5	2	2.2	0	✓
freire2	1	4	35	26.0	3	5.1	1	✓
cohencb	1	5	31	23.6	4	4.2	1	✓
egcd1	1	8	108	43.1	1	12.8	8	—
egcd2	2	10	209	60.8	8	14.6	12	✓
egcd3	3	12	475	67.0	14	23.4	25	✓
lcm1	3	6	203	38.9	12	14.2	0	✓
lcm2	1	6	52	14.9	1	0.9	10	✓
prodbin	1	5	61	28.3	3	1.1	10	—
prod4br	1	6	42	9.6	4	8.6	7	✓
fermat1	3	5	217	75.7	6	6.2	1	✓
fermat2	1	5	70	25.8	2	5.2	0	✓
knuth	1	8	113	57.1	4	24.6	6	✓
geo1	1	4	25	16.7	2	1.5	4	✓
geo2	1	4	45	24.1	1	2.1	10	✓
geo3	1	5	65	22.1	1	2.7	12	✓
ps2	1	3	25	21.1	2	4.0	0	✓
ps3	1	3	25	21.9	2	4.2	0	✓
ps4	1	3	25	23.5	2	4.9	0	✓
ps5	1	3	24	24.9	2	7.4	0	✓
ps6	1	3	25	25.0	2	69.5	0	✓
<b>total</b>			2392	825.9	118	149	266.3	<b>22/27</b>

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ps4	1	3	25	23.5	2	4.9	0	✓
ps5	1	3	24	24.9	2	7.4	0	✓
ps6	1	3	25	25.0	2	69.5	0	✓
<b>total</b>			2392	825.9	118	149	266.3	<b>22/27</b>

# Summary

Hybrid invariant generation:

- **DIG** employs geometric concepts for dynamic invariant inference
  - Build *max-plus polyhedra* to find disjunctive polynomial invariants
  - Combine max and min relations for expressive power
  - Introduce new classes of *weak* max/min invariants that retain expressiveness and have polynomial time complexity
- **KIP** verifies invariants using program code (k-induction, SMT solving, lemmas learning, redundancy elimination, parallelism)

Results:

- Identify strong enough invariants to prove correctness of 36/41 programs
- Produce *no spurious* results
- Take 2 minutes per program, on avg, to find and prove invs

# Thank you for your attention !

Project (DIG + KIP) is open source and available at

<http://cs.unm.edu/~tnguyen>

Ask me about (or check my webpage for details)

- Dynamic analysis on complex array invariants (identified 60% of array relations in AES)
- Static analysis on octagonal and max-plus invariants using quantifier elimination
- Automatic program repair using test-input generation (equivalence theorem: program reachability  $\equiv$  program synthesis)



# Combining Max and Min-plus Invariants

## Combining Max and Min-plus Invariants

```
def ex2(x):  
    if x ≥ 0:  
        y = x + 1  
    else:  
        y = x - 1  
    b = y > 10  
    [L]  
    return b
```

x	y	b
-50	-51	0
-33	-34	0
9	10	0
10	11	1
12	13	1
40	41	1

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- By building max and min-plus polyhedra in 3D, DIG obtains

$$1 \geq b \geq 0, \max(y - 10, 0) \geq b, b + 10 \geq \min(y, 11),$$

i.e.,

$$1 \geq b \geq 0 \wedge \max(y - 10, 0) \geq b \Rightarrow b = 0 \Rightarrow y \leq 10$$

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- Logically equivalent to the iff condition

$$b = 0 \Leftrightarrow y \leq 10$$