## Compositional Static Race Detection at Scale, without False Positives

Ilya Sergey





Joint work with

Sam Blackshear, Peter O'Hearn, Nikos Gorogiannis



## Key Messages

Static analyses for concurrency can be made scalable and precise.

Unsound (and incomplete) static analyses can be *principled*, satisfying meaningful theorems.

One can have an unsound but effective static analysis, which has significant *industrial impact*, and which is supported by a *meaningful theorem*.

#### Practice

#### **RACERD: Compositional Static Race Detection**

SAM BLACKSHEAR, Facebook, USA
NIKOS GOROGIANNIS, Facebook, UK and Middlesex University London, UK
PETER W. O'HEARN, Facebook, UK and University College London, UK
ILYA SERGEY\*, Yale-NUS College, Singapore and University College London, UK

Automatic static detection of data races is one of the most basic problems in reasoning about concurrency. We present RacerD—a static program analysis for detecting data races in Java programs which is fast, can scale to large code, and has proven effective in an industrial software engineering scenario. To our knowledge, RacerD is the first inter-procedural, compositional data race detector which has been empirically shown to have non-trivial precision and impact. Due to its compositionality, it can analyze code changes quickly, and this allows it to perform *continuous reasoning* about a large, rapidly changing codebase as part of deployment within a continuous integration ecosystem. In contrast to previous static race detectors, its design favors reporting high-confidence bugs over ensuring their absence. RacerD has been in deployment for over a year at Facebook, where it has flagged over 2500 issues that have been fixed by developers before reaching production. It has been important in enabling the development of new code as well as fixing old code: it helped support the conversion of part of the main Facebook Android app from a single-threaded to a multi-threaded architecture. In this paper we describe RacerD's design, implementation, deployment and impact.

CCS Concepts: • Theory of computation  $\rightarrow$  Program analysis; • Software and its engineering  $\rightarrow$  Concurrent programming structures;

Additional Key Words and Phrases: Concurrency, Static Analysis, Race Freedom

#### **ACM Reference Format:**

Sam Blackshear, Nikos Gorogiannis, Peter W. O'Hearn, and Ilya Sergey. 2018. RACERD: Compositional Static Race Detection. *Proc. ACM Program. Lang.* 2, OOPSLA, Article 144 (November 2018), 28 pages. https://doi.org/10.1145/3276514

00PSLA'18

#### Theory





#### A True Positives Theorem for a Static Race Detector

NIKOS GOROGIANNIS, Facebook, UK and Middlesex University London, UK
PETER W. O'HEARN, Facebook, UK and University College London, UK
ILYA SERGEY\*, Yale-NUS College, Singapore and National University of Singapore, Singapore

RACERD is a static race detector that has been proven to be effective in engineering practice: it has seen thousands of data races fixed by developers before reaching production, and has supported the migration of Facebook's Android app rendering infrastructure from a single-threaded to a multi-threaded architecture. We prove a True Positives Theorem stating that, under certain assumptions, an idealized theoretical version of the analysis *never reports a false positive*. We also provide an empirical evaluation of an implementation of this analysis, versus the original RACERD.

The theorem was motivated in the first case by the desire to understand the observation from production that RACERD was providing remarkably accurate signal to developers, and then the theorem guided further analyzer design decisions. Technically, our result can be seen as saying that the analysis computes an underapproximation of an over-approximation, which is the reverse of the more usual (over of under) situation in static analysis. Until now, static analyzers that are effective in practice but unsound have often been regarded as ad hoc; in contrast, we suggest that, in the future, theorems of this variety might be generally useful in understanding, justifying and designing effective static analyses for bug catching.

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Additional Key Words and Phrases: Concurrency, Static Analysis, Race Freedom, Abstract Interpretation

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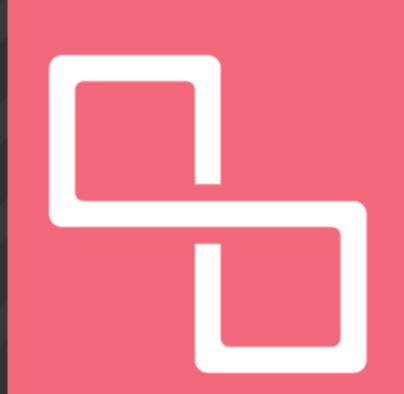
Nikos Gorogiannis, Peter W. O'Hearn, and Ilya Sergey. 2019. A True Positives Theorem for a Static Race Detector. *Proc. ACM Program. Lang.* 3, POPL, Article 57 (January 2019), 29 pages. https://doi.org/10.1145/3290370

POPL'19

#### Part

#### RacerD:

Compositional Static Race Detection



#### Litho: A declarative UI framework for Android

**GET STARTED** 

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**TUTORIAL** 

#### Litho Component

Fetch data

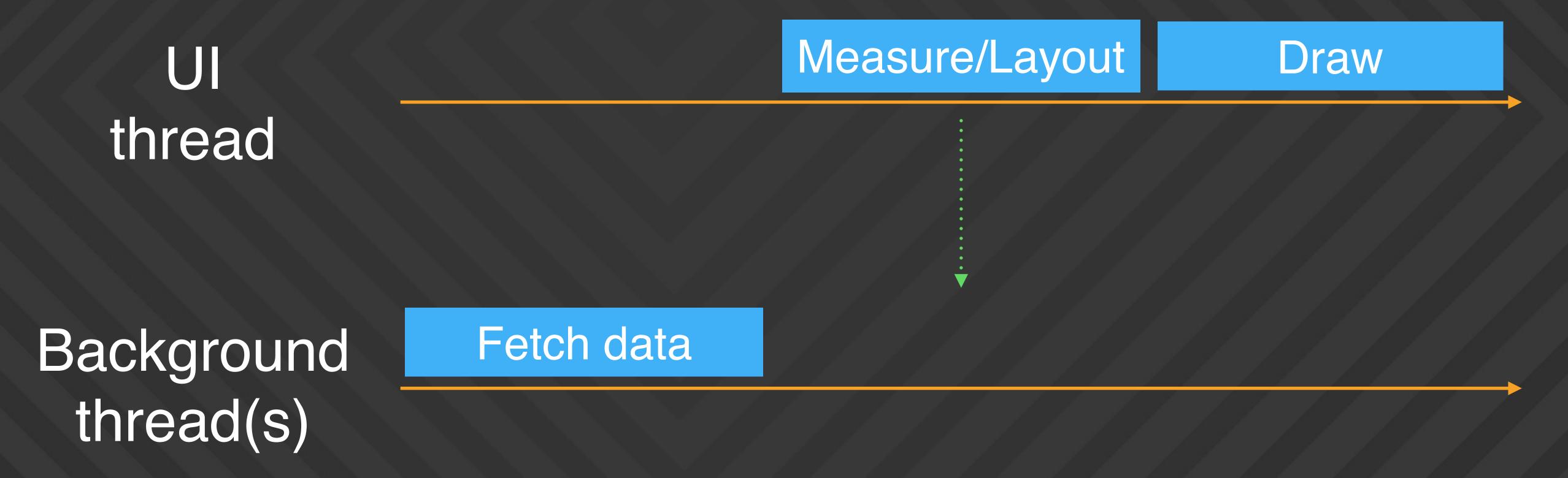
Talk to network

Measure/Layout Determine size and position !

Draw

Render and attach

#### Moving layout to background for better perf



BUT: to migrate, Measure/Layout step needs to be thread-safe. Otherwise...

#### Adding concurrency can introduce data races

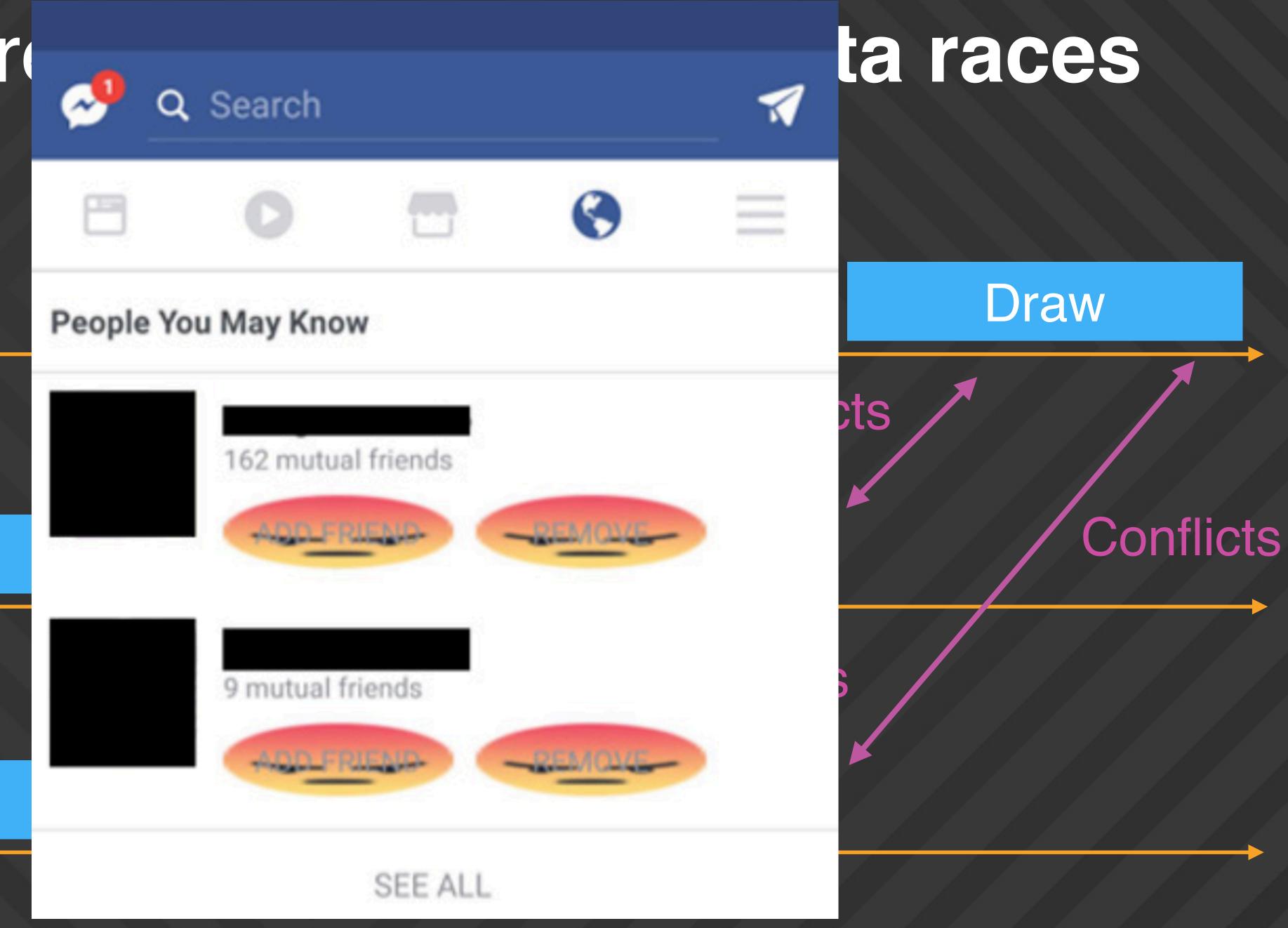
# Data race: two concurrent accesses to the same memory location where at least one is a write.

#### Concurr

UI thread

Background thread 1

Background thread 2



#### Adding concurrency to sequential code is scary

Problem 1: 1000s of existing components. Where should we add synchronization to avoid races?

Problem 2: Nondeterminism makes it hard to test for races. How do we prevent regressions?

Static race detector can show us where to add synchronization + prevent regressions at code review time.

#### Devs need static analysis for migration

Talking with \_\_\_\_\_, one of our managers - we realized that the timeline of background layout in feed might be closely tied to the timeline of static analysis - I'm wondering if you have your roadmap already fleshed out.

#### Stringent requirements for helpful analysis

Interprocedural

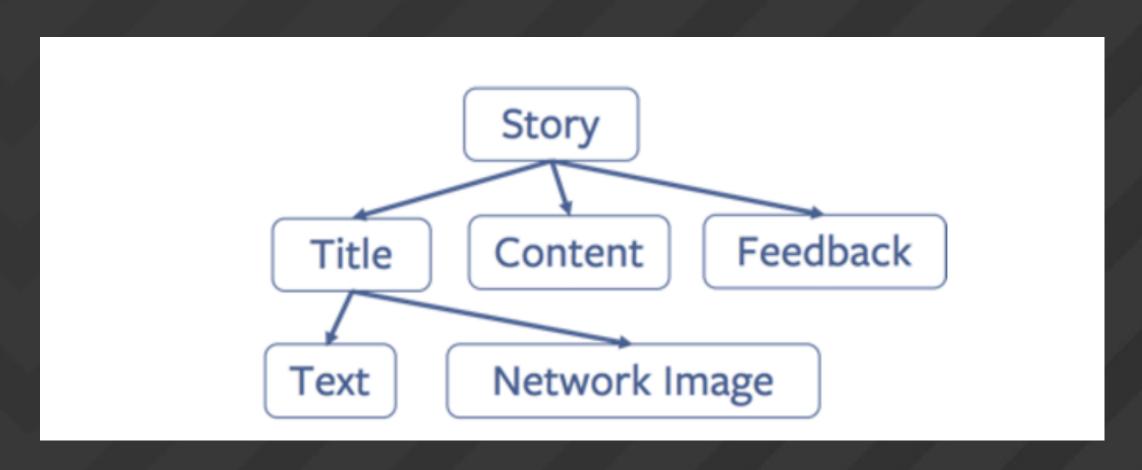
Scalable + incremental

Low annotation burden

High signal >> catching all bugs

will the eventual thread safe annotation be recursive? Will it check that dependencies, at least how they're used, are thread safe?

Like · Reply · Share · ② 2 · October 14, 2016 at 11:04pm



@GuardedBy("this") Classname object;

#### RacerD: compositional static race detection

(1) Motivation: assist Litho migration + detect regressions

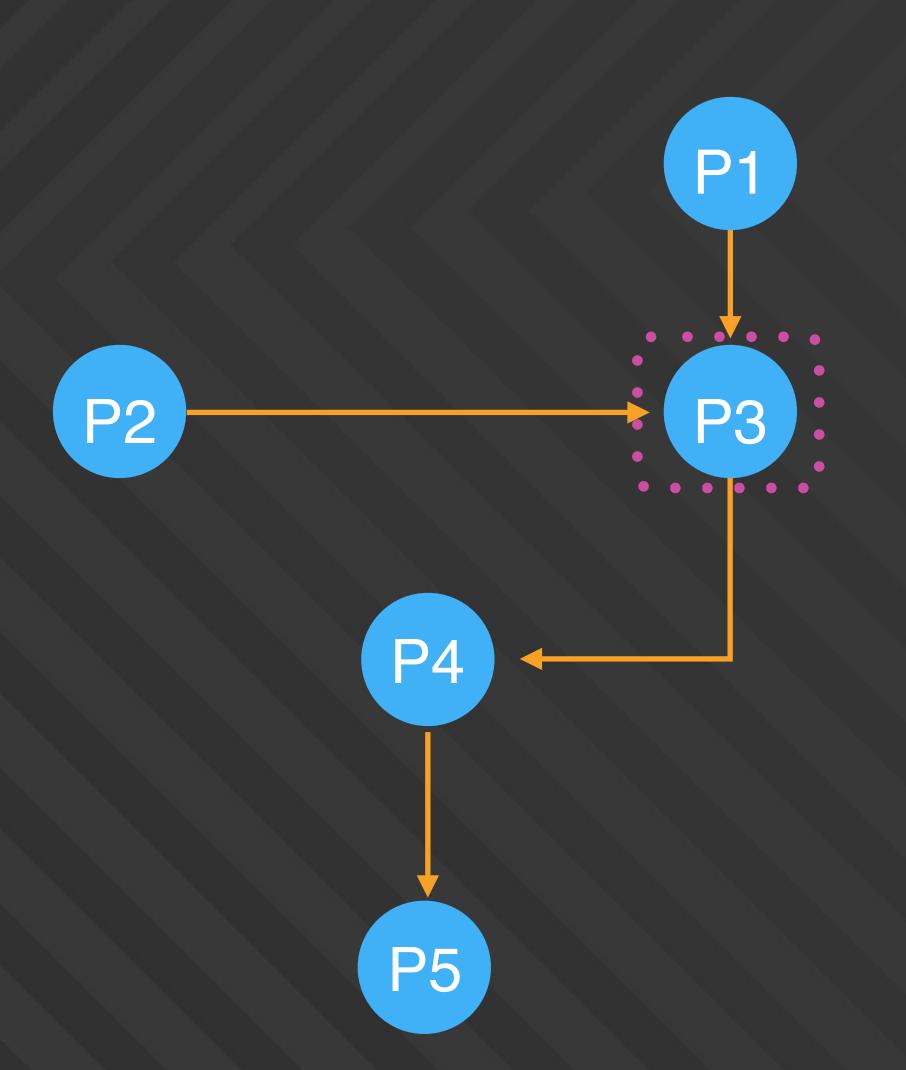
(2) RacerD deep dive: design, domains, reporting

(3) Evaluation: RacerD vs static/dynamic race detectors, RacerD @ FB

#### RacerD Design Principles

- Be compositional; don't do whole-program analysis
- Report races between syntactically identical access paths;
   don't attempt a general alias analysis
- Reason sequentially about memory accesses, locks, and threads; don't explore interleaving
- Occam's razor; don't use complex techniques (unless forced)

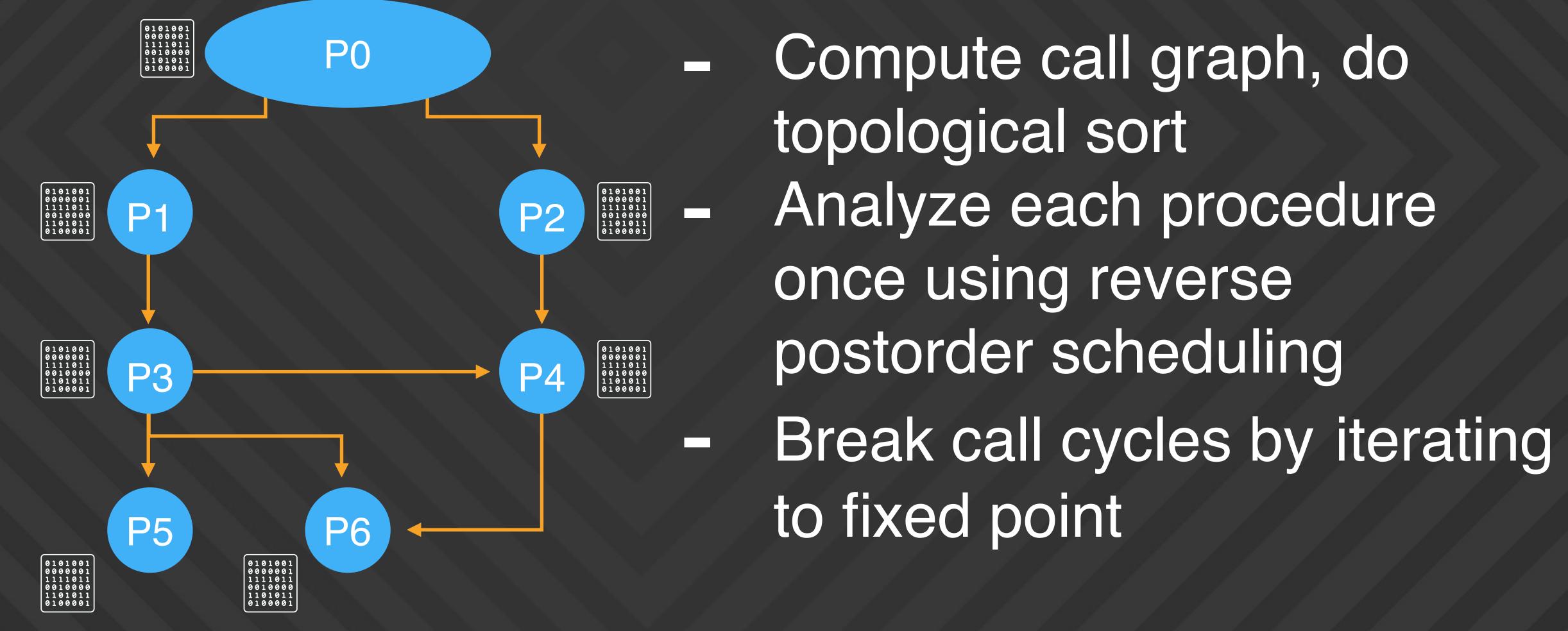
#### Background: compositional analysis



When analyzing P3:

- Will have summary for callee P4
- But don't know anything about callers P1, P2, or transitive callee P5
- Need to compute summary for P3 usable in any calling context

#### Background: compositional analysis



Scalable: analyze each procedure just once (without cycles)

#### Computing procedure summaries



get and reset access same memory location reset performs a write under synchronization get uses no synchronization

```
class Counter {
  private int mCount;
  int get() {
    return this.mCount;
  private void set(int i) {
    this.mCount = i;
  synchronized void reset() {
    set(0);
```

#### RacerD abstract domains

- (1) Can access touch the same memory?
- (2) Can accesses happen concurrently?

access snapshots

$$A \subseteq \mathcal{A} = \{ \langle \pi, k, \ell, t, o \rangle \mid k \in \{ \mathsf{rd}, \mathsf{wr} \} \}$$

(1) access paths

$$\pi \in \mathsf{Path} = \mathsf{Var} \times \mathsf{Field}^*$$

(2) lock count

$$\ell \in \mathcal{L} = \mathbb{N}^{\top}$$

(2) concurrent threads

$$t \in \mathcal{T}$$
 = NoThread  $\Box$  AnyThreadButMain  $\Box$  AnyThread

ownership value  $o \in \mathcal{O}$  ownership environment  $E \subseteq \mathcal{E}$ 

$$o \in \mathcal{O}$$
 = OwnedIf( $\wp(\mathbb{N})$ )  $\square$  Unowned

 $E \subseteq \mathcal{E}$  = Path  $\to \mathcal{O}$ 

domain

$$d \in \mathcal{D} = \mathcal{L} \times \mathcal{T} \times \mathcal{E} \times \mathcal{A}$$

#### Concurrent context inference

Accesses are NoThread unless we see:

Android thread assertions
Introduce AnyThreadButMain

```
void m() {
  ThreadUtils.assertMainThread();
  this.f = ...
}
```

java.util.concurrent annots

```
@ThreadSafe class C { ...
```

Lock usage Introduce AnyThread

```
synchronized void m() { this.f = ... }
```

#### Ownership identifies clearly safe accesses

```
Obj local = new Obj();
local.f = ... // safe write
```

Freshly allocated ownership

```
Obj objFactory() {
  return new Obj();
Obj local = objFactory();
          ... // safe write
```

Returning ownership

#### Ownership can be conditional

```
private void writeF(Obj a) {
  a.f = ...
}
Obj o = new Obj();
writeF(o); // safe
```

Safe if formal is owned by caller

```
Builder setX(X x) {
   this.x = x;
   return this;
}

new Builder() setX(x) setY(y); // safe
global.set(X).f = /; // not safe
```

Returns ownership if formal is owned by caller

#### Ownership preconditions in summaries

```
private void writeF(Obj a) {
  a.f = ...
}
summ: { a.f if ¬owned(a) }
```

```
Obj o = new Obj();
owned(o)
writeF(o);
owned(o) \lambda { o.f if ¬owned(o) }
owned(o) \lambda {}
```

#### Ownership post can depend on ownership pre

```
Builder setX(X x) {
   this.x = x;
   return this;
}
summ: { (this.x if ¬owned(this) } ^
        owned(ret) if owned(this)
```

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domain

$$d \in \mathcal{D} = \mathcal{L} \times \mathcal{T} \times \mathcal{E} \times \mathcal{A}$$

#### Reporting:

Can  $\langle \pi_1, o_1, \ell_1, t_1, k_1 \rangle$  and  $\langle \pi_2, o_2, \ell_2, t_2, k_2 \rangle$  race?

### two concurrent accesses to the same memory location where at least one is a write.

- $\begin{array}{l} (1) \ \pi_1 = \pi_2 \\ (2) \ o_1 \neq \mathsf{Owned} \ \land \ o_2 \neq \mathsf{Owned} \end{array}$
- $(3) \ \ell_1 = 0 \ \lor \ \ell_2 = 0$
- (4)  $t_1 \sqcup t_2 = \mathsf{AnyThread}$
- (5)  $k_1 = \text{wr} \lor k_2 = \text{wr}$

#### Applying summaries: see paper

$$\operatorname{apply}(\ell_{caller},\ell_{callee}) \triangleq \ell_{caller} +^{\top}\ell_{callee} \tag{A}$$

$$\operatorname{apply}(\ell_{caller},t_{callee}) \triangleq \begin{cases} \operatorname{AnyThreadButMain} & \text{if } t_{callee} = \operatorname{AnyThreadButMain} \\ t_{caller} & \text{otherwise} \end{cases} \tag{B}$$

$$E(e) \triangleq \begin{cases} \operatorname{O} & \text{if } \mathbf{e} = \pi \text{ and } (\pi,o) \in E \\ \operatorname{Unowned} & \text{if } \mathbf{e} = \pi \text{ and } \pi \notin \operatorname{dom}(E) \\ \operatorname{OwnedIf}(\emptyset) & \text{otherwise} \end{cases} \tag{C}$$

$$\operatorname{apply}(E_{caller},\vec{\mathbf{e}},o_{callee}) \triangleq \begin{cases} \operatorname{Unowned} & \text{if } o_{callee} = \operatorname{Unowned} \\ \operatorname{OwnedIf}(\emptyset) & \text{if } o_{callee} = \operatorname{OwnedIf}(\emptyset) \\ \sqcup_{i \in N} E_{caller}(\mathbf{e}_i) & \text{if } o_{callee} = \operatorname{OwnedIf}(N) \end{cases} \tag{D}$$

$$\operatorname{apply}(E_{caller},\vec{\mathbf{e}},\pi_{ret},E_{callee}) \triangleq E_{caller}[\pi_{ret} \mapsto \operatorname{apply}(E_{caller},E_{callee}(ret),\vec{\mathbf{e}})] \tag{E}$$

$$\operatorname{subst}(\pi,\vec{\mathbf{e}}) \triangleq \begin{cases} y.f_1.\cdots.f_n.g_1.\cdots.g_m & \text{if } \begin{cases} \pi = x.g_1.\cdots.g_m, \\ x \text{ is the } i\text{th formal, and } \\ e_i = y.f_1.\cdots.f_n \\ \text{otherwise} \end{cases} \end{cases}$$

$$\operatorname{apply}(A_{caller},\ell_{caller},E_{caller},E_{callee}) \triangleq A_{caller} \sqcup \begin{cases} \langle \pi',k,\ell',t',o' \rangle & \pi' = \operatorname{subst}(\pi,\vec{\mathbf{e}}), \\ \ell' = \operatorname{apply}(\ell_{caller},\ell), \\ \ell' = \operatorname{apply}(\ell_{caller},\ell), \\ \ell' = \operatorname{apply}(E_{caller},\ell), \\ \ell$$

Fig. 5. Applying a callee summary at a call site  $\pi_{ret} := \text{call } m(\vec{e})$ .

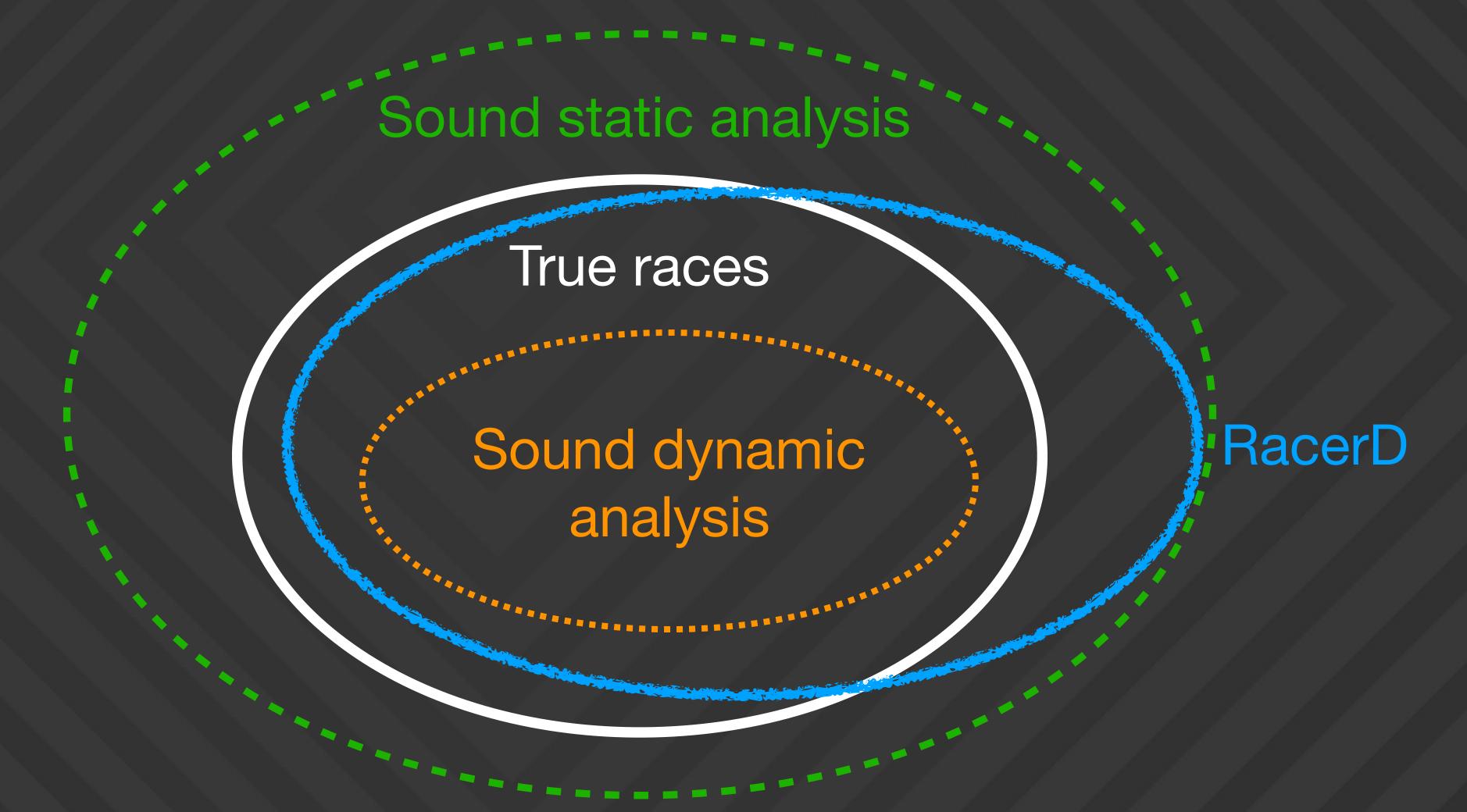
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#### RacerD vs Static and Dynamic Analysis tools



Want: more true races than dynamic, fewer false positives than static

#### RacerD vs Chord (static)

- Ran RacerD on DaCapo benchmarks used by Chord (orig. PLDI'06) in PLDI'18 paper
- RacerD reports >3X fewer alarms in all cases,
   >10x fewer on half of the benchmarks
- RacerD found all true bugs reported by Chord (+ more!)
- Time: 0.6s 25s (RacerD) vs 1m 1h 47m (Chord)

#### RacerD vs Chord (static)

Program	# Files	# LOC	CHORD		RACERD		Alarm
			# Alarms	Runtime	# Alarms	Runtime	Overlap
hedc	133	11,767	152	4m 47s	49	1.5s	11%
ftp	140	12,050	522	5m 14s	39	1.5s	20%
weblech	12	1,309	30	11m 02s	9	0.6s	63%
jspider	214	7,413	257	1m 54s	13	1.4s	10%
avrora	470	68,864	966	8m 49s	81	7s	18%
luindex	331	36,151	940	3m 26s	183	5s	64%
sunflow	170	21,960	958	42m 44s	43	2.8s	49%
xalan	975	175,784	1870	1h 47m	421	21.5s	38%

#### RacerD vs DroidRacer (dynamic)

- Ran RacerD on DaCapo benchmarks used by DroidRacer in PLDI'14 paper
- RacerD found more true bugs than DroidRacer on each benchmark
- In two cases, also reported fewer false positives
- Time: 12s 3m (RacerD) vs ""few seconds to few hours (DroidTacer)

# RacerD reports fewer false positives than sound static tools and finds more bugs than dynamic tools.

#### Finding data race regressions Impact (1y)

~500

PROGRAMMERS REACHED

~7K ~4K

REPORTS

FIXES



#### RacerD made migration faster/easier/safer

 100s of Litho components safely migrated to background layout in <6 mo by 2 engineers</li>

- 1000+ potential races fixed during migration

 Only 3 false negatives reported by engineers (all analyzer bugs that we fixed)

## Engineer Comments

better. The thread safety violations are doubly useful - since these help catch nasty and hard to debug bugs that can commonly happen in our multi-thread UI stack on Android

Infer was really instrumental in ensuring thread safety in Litho code. This allowed us to ship Newsfeed layout on a background thread and get huge wins in terms of scroll performance in FB4a

Without Infer, multithreading in News Feed would not have been tenable

#### RacerD Limitations

- 1. Races on aliased, syntactically distinct access paths not caugh
  - -- Tried alias analysis, too many FPs. Couldn't deploy.
- 2. Misses races on local objects that escape, "deep" ownership analysis
  - -- Tried escape analysis, too many FPs. Couldn't deploy.
- 3. Not fully underapproximate
  - -- Tried join-as-meet, but eliminates too many valuable reports. Couldn't deploy
- 4. Works for simple mutex locks, but not subtler constructs like R/W locks
- 5. No reasoning about weak memory, volatile, or other fine-grained concurrency
- 6. No soundness/completeness theorem

#### Helps real programmers write concurrent code anyway

## Try RacerD

https://fbinfer.com/docs/racerd.html

or Google "Facebook RacerD"

# Checkpoint

- (1) **RacerD**: scalable + low annotation static race detection designed around completeness rather than soundness
- (2) Detected 1000s of bugs in prod at FB + enabled Litho migration
- (3) Can we have a "True Positives Theorem"?

#### Part 2

A True Positives Theorem for a Static Race Detector

#### Context

- 1. We had a demonstrably-effective industrial analysis: RacerD (OOPSLA'18); >3k fixes in Facebook Java
- 2. No soundness theorem
- 3. Architecture: compositional abstract interpreter
- 4. No heuristic alarm filtering

#### Just ad hoc?



#### Our reaction:

Semantics/theory should understand/explain, not lecture.

#### RACERD: Compositional Static Race Detection

OOPSTA 18 SAM BLACKSHEAR, Facebook, USA NIKOS GOROGIANNIS, Facebook, UK and Middlesex University London, UK PETER W. O'HEARN, Facebook, UK and University College London, UK ILYA SERGEY\*, Yale-NUS College, Singapore and University College London, UK

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In theory ...

Unsound for verification (misses bugs)

Unsound for testing (reports non-bugs)

yet, over 6 months of use and 2500 bugs reported

Only 3 known races missed

Less than 20% false positives

Can we provably make it 0%?

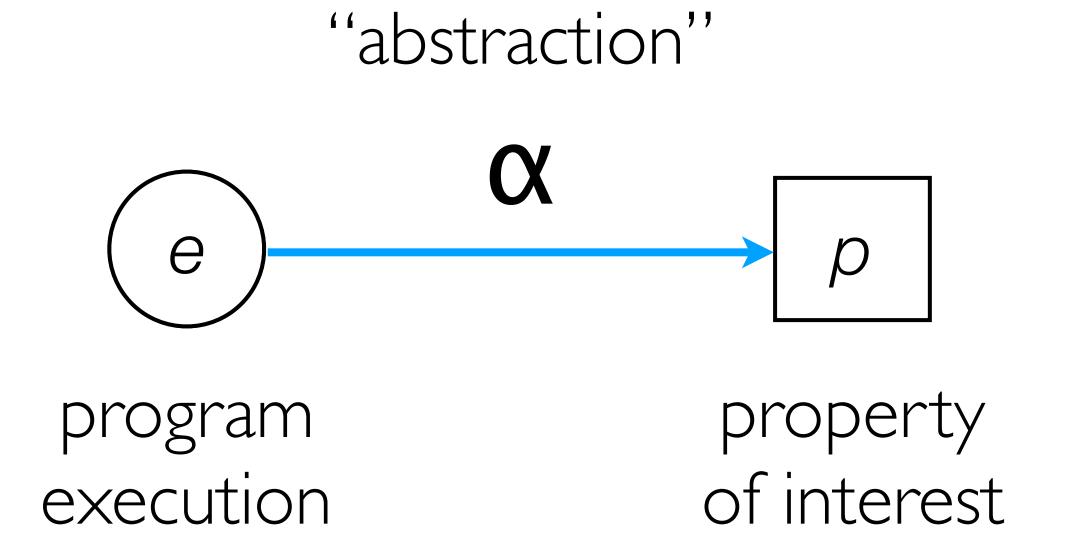
### Conjecture

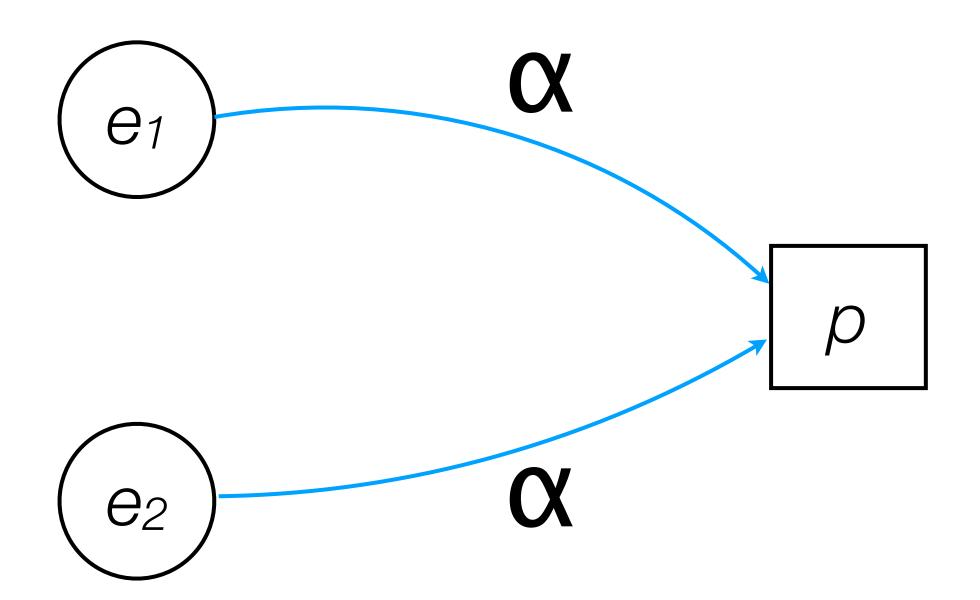
#### **True Positives Theorem:**

Under certain assumptions, the static bug detector reports no false positives.

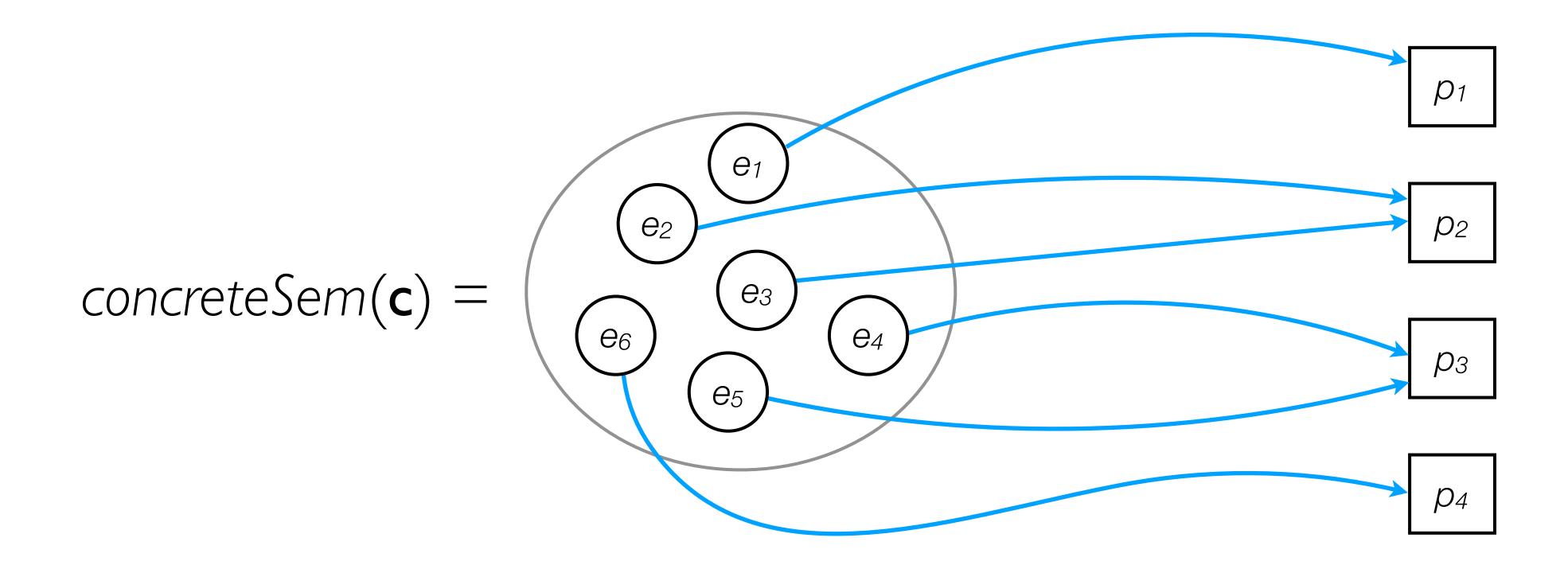
# Static Analyses for Program Validation

#### The Essence of Static Analysis

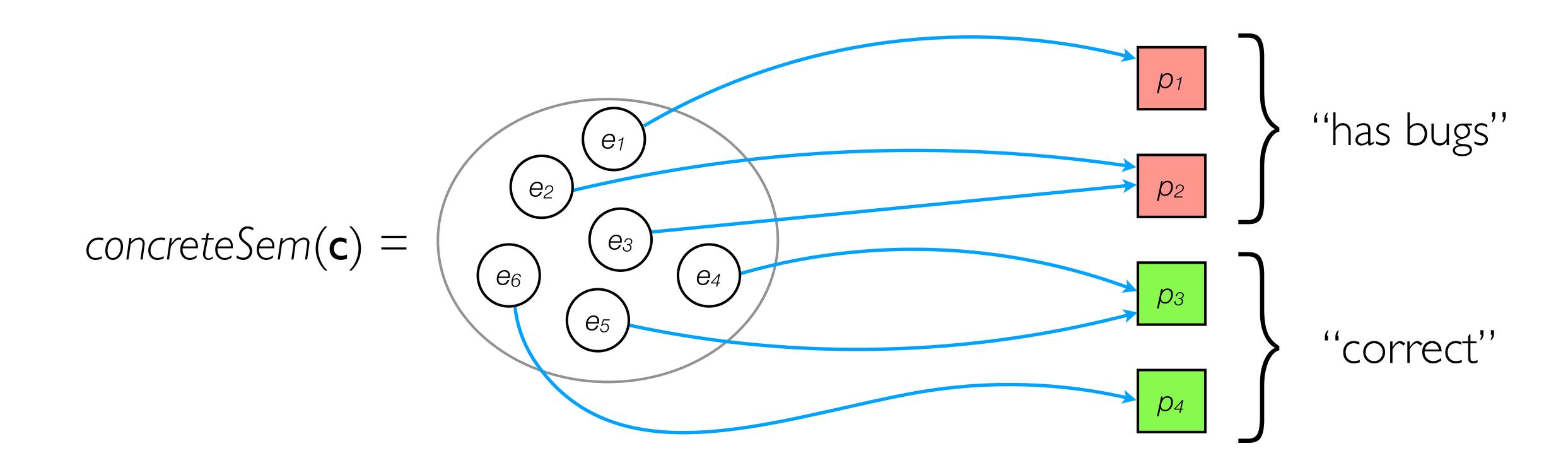




#### Static Analysis

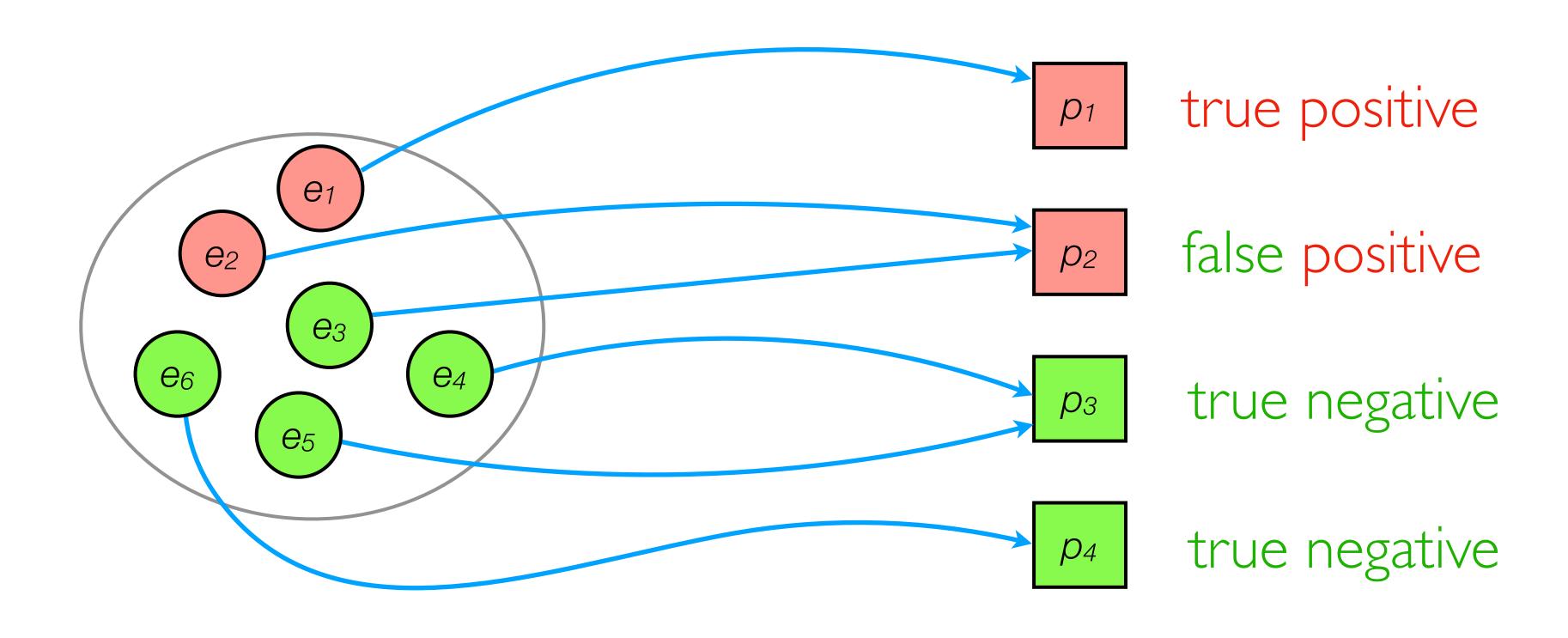


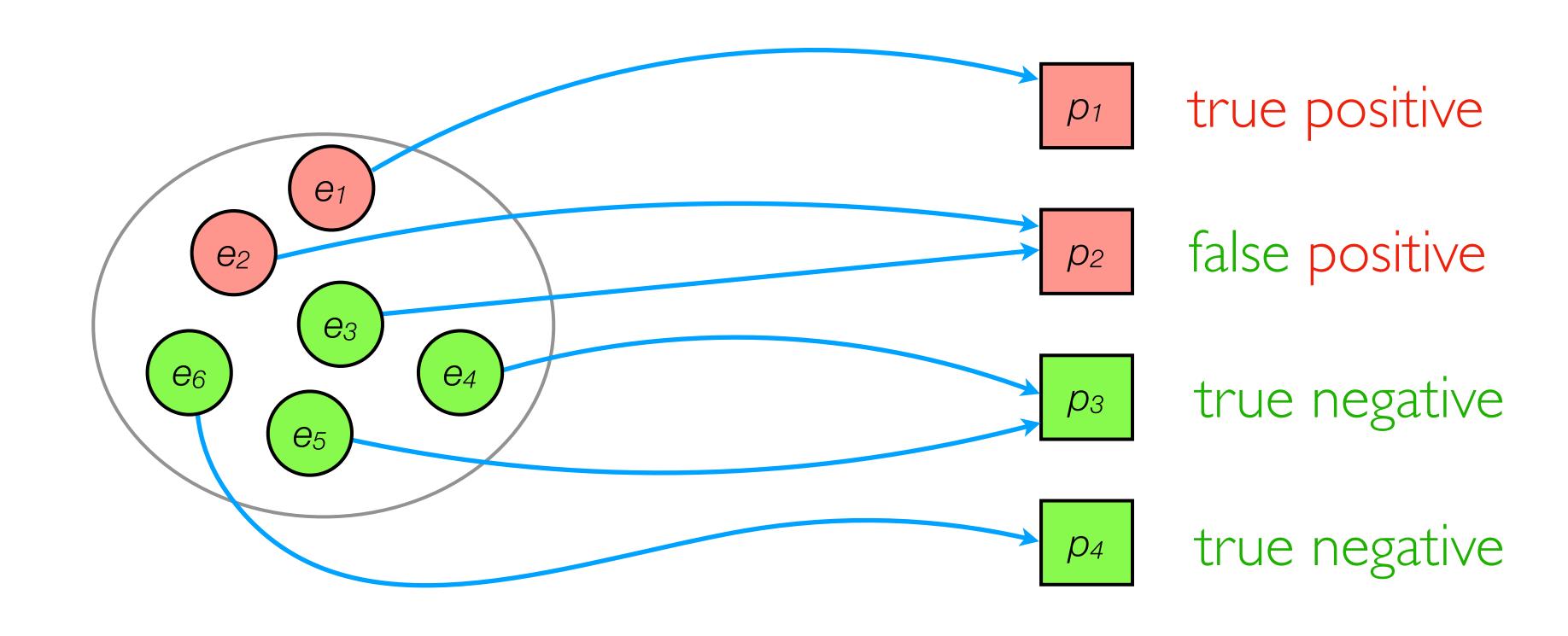
### Static Analysis

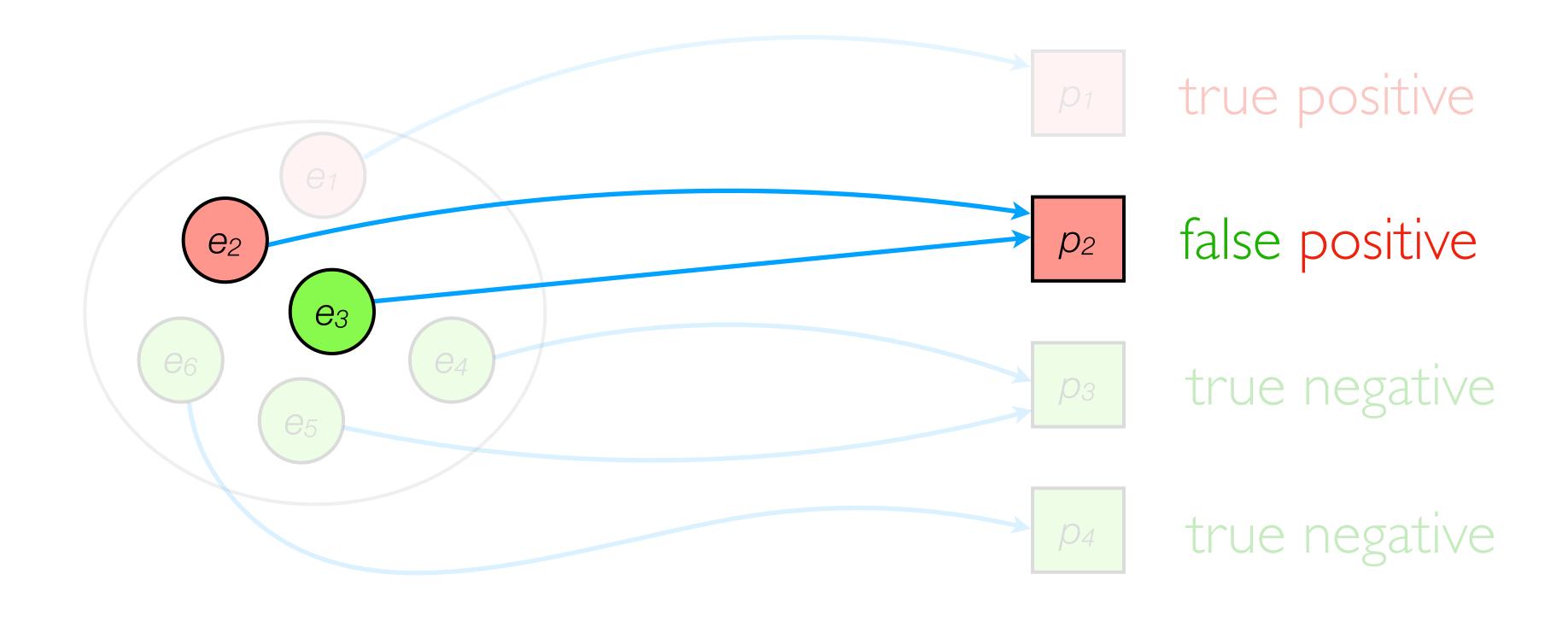


# Verifier or a Bug Detector?

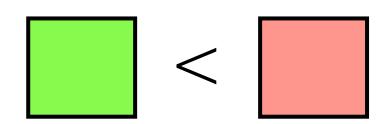
# Program Verifier

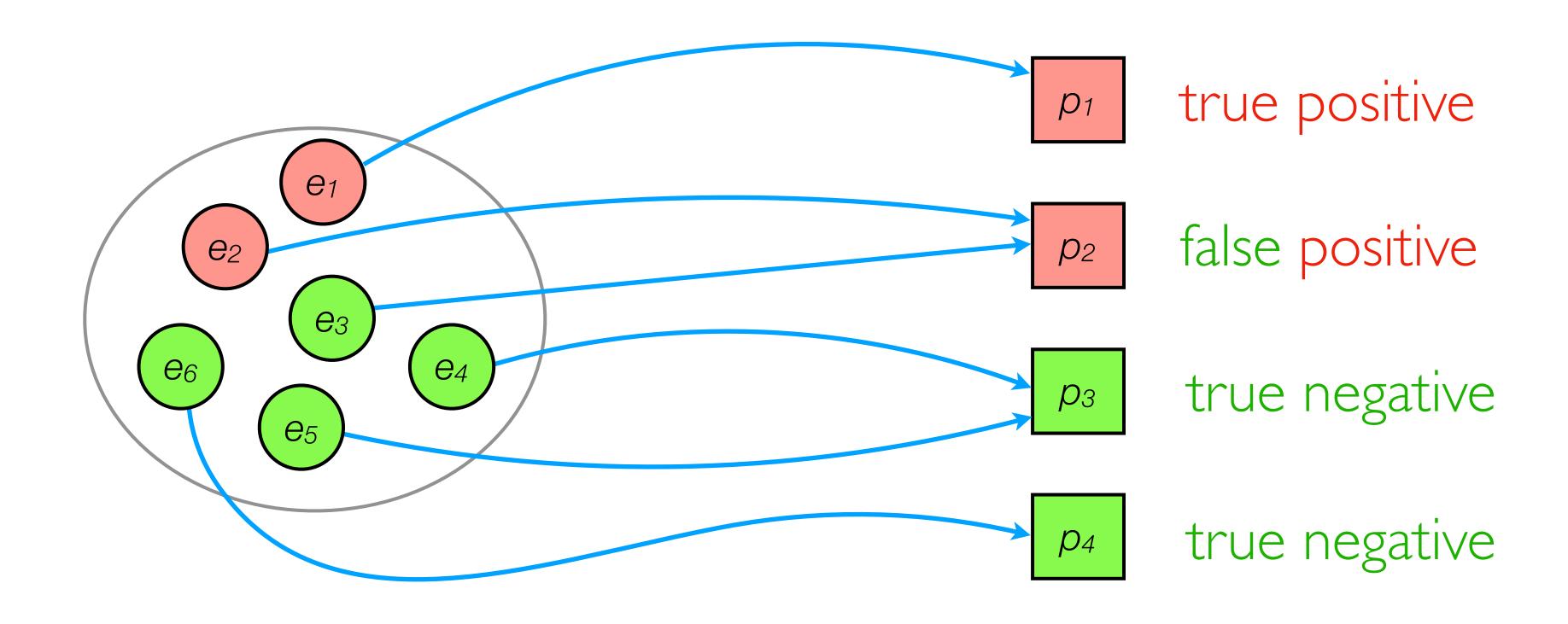




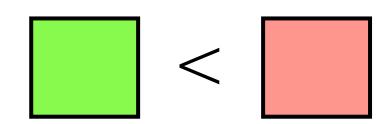


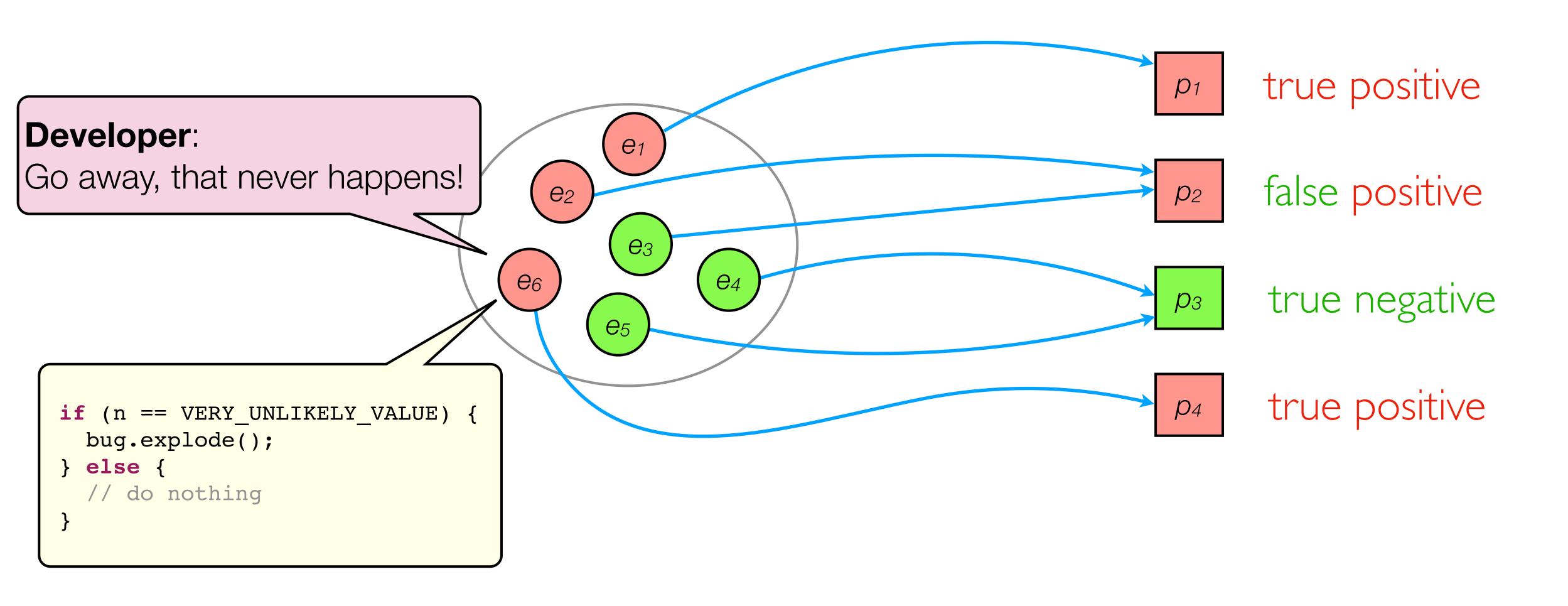
abstract over-approximation



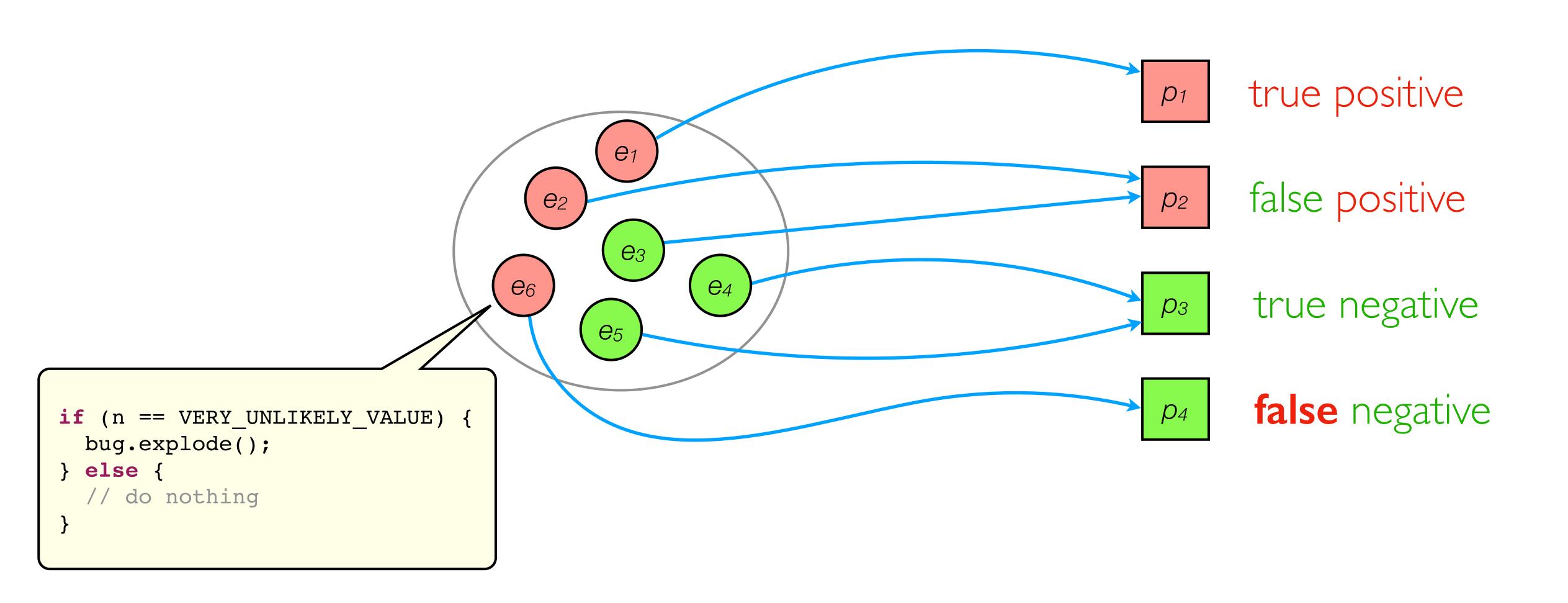


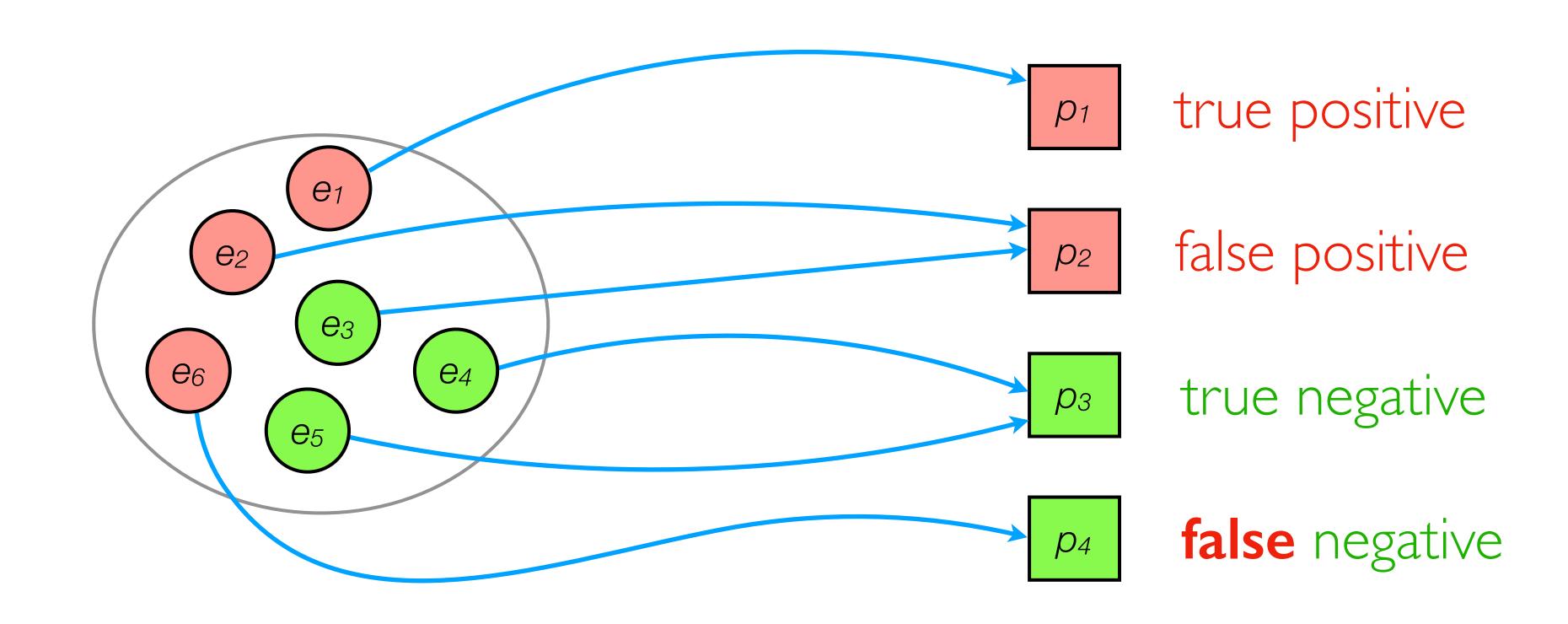
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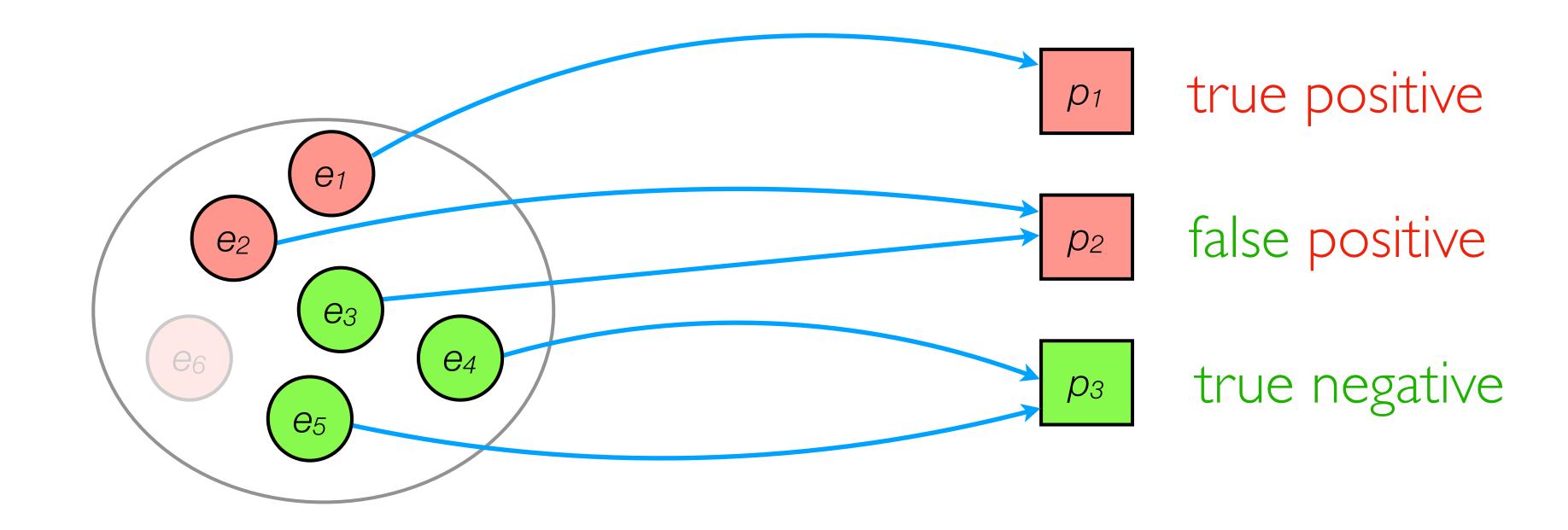




#### Unsound Program "Verifier"

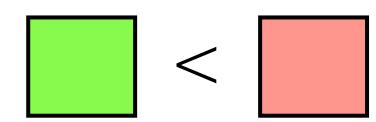






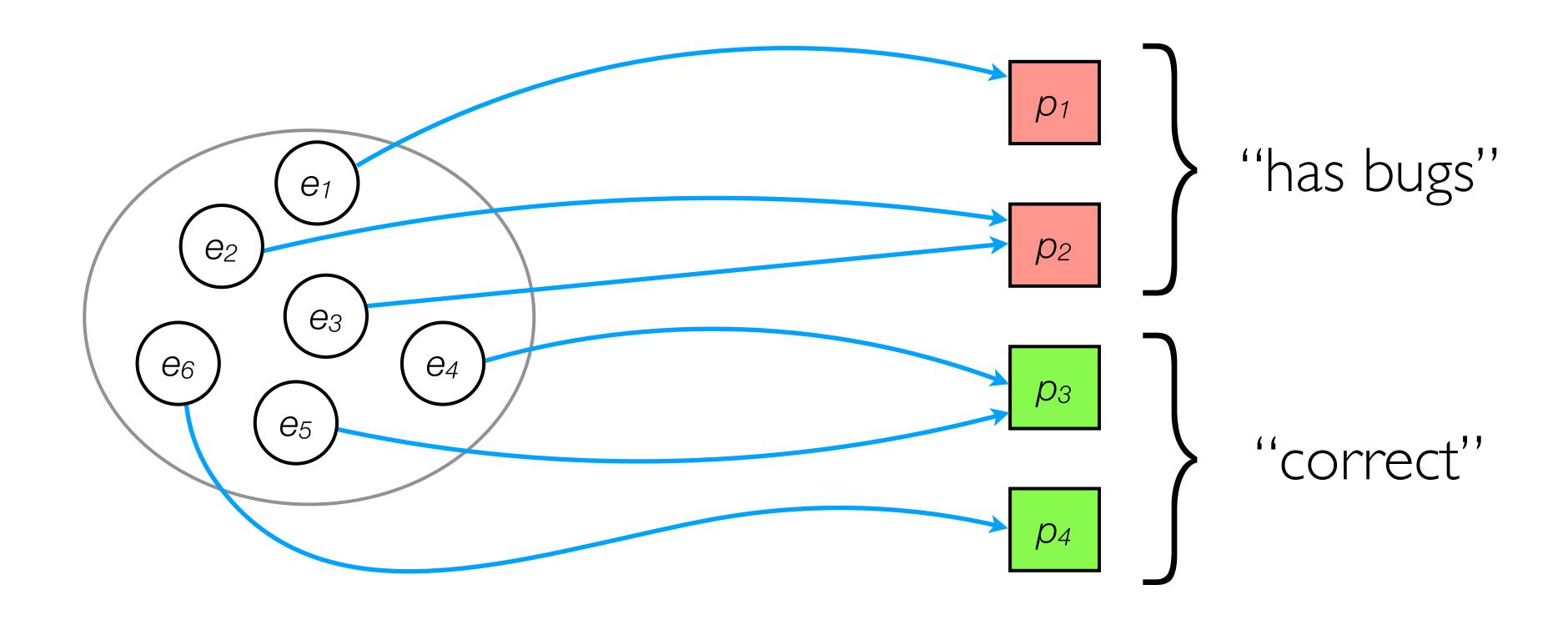
concrete under-approximation

abstract over-approximation

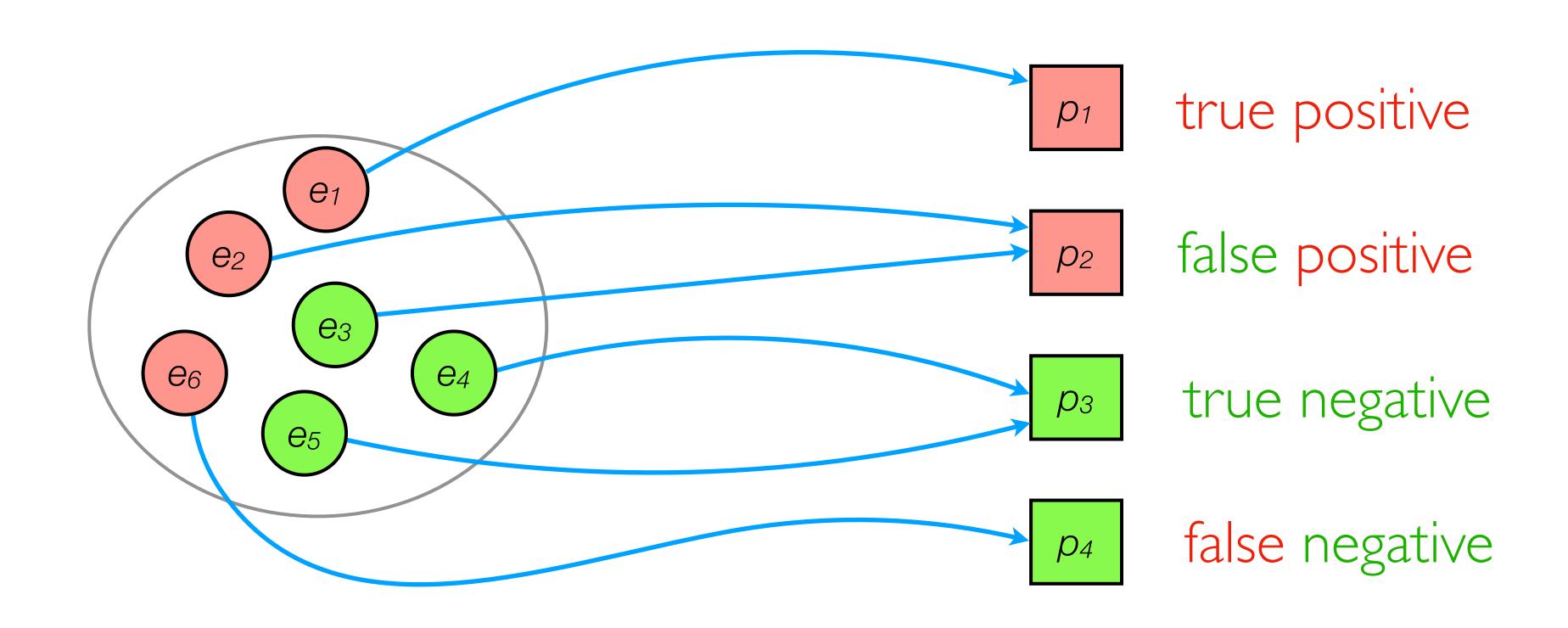


#### Sound Static Verifiers

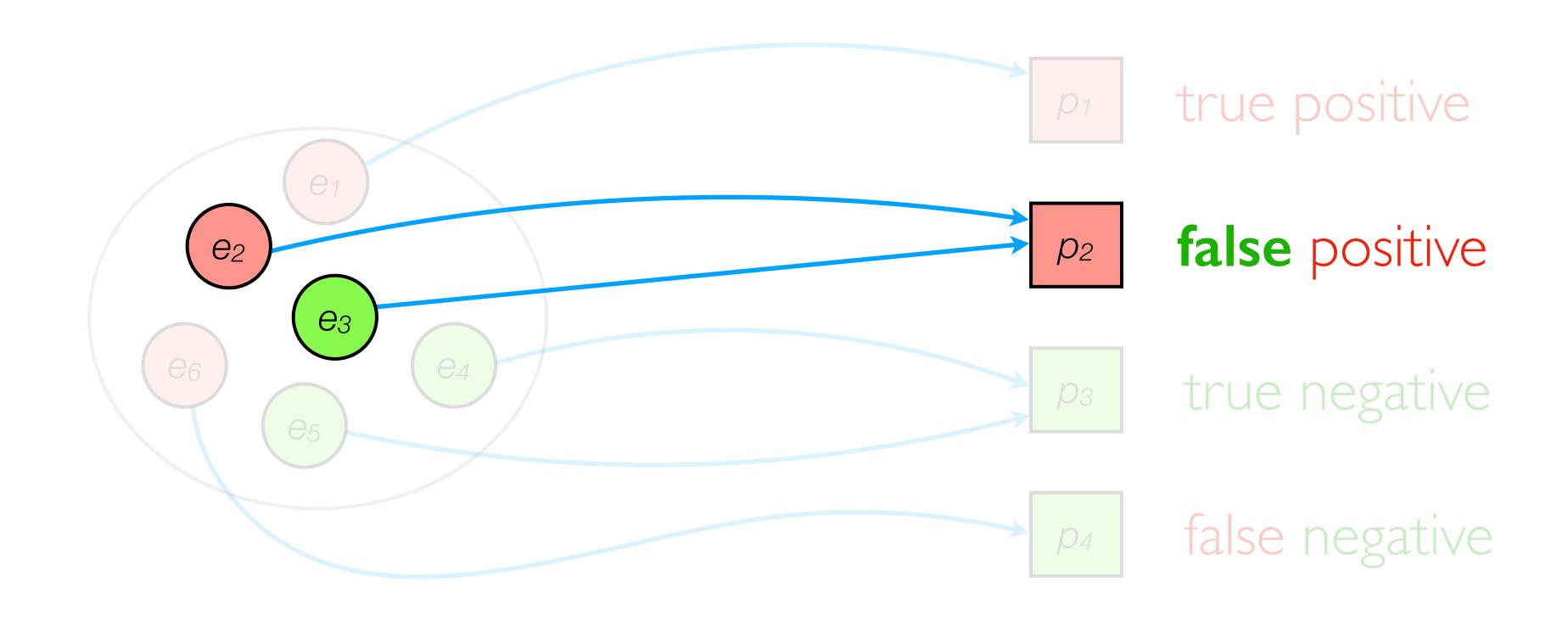
- False negatives (bugs missed) are bad
- False positives (non-bugs reported) are okay
- Constructed as over-approximation (of under-approximation)
- Soundness Theorem:
  Under certain assumptions about the programs, the analyser has no false negatives.



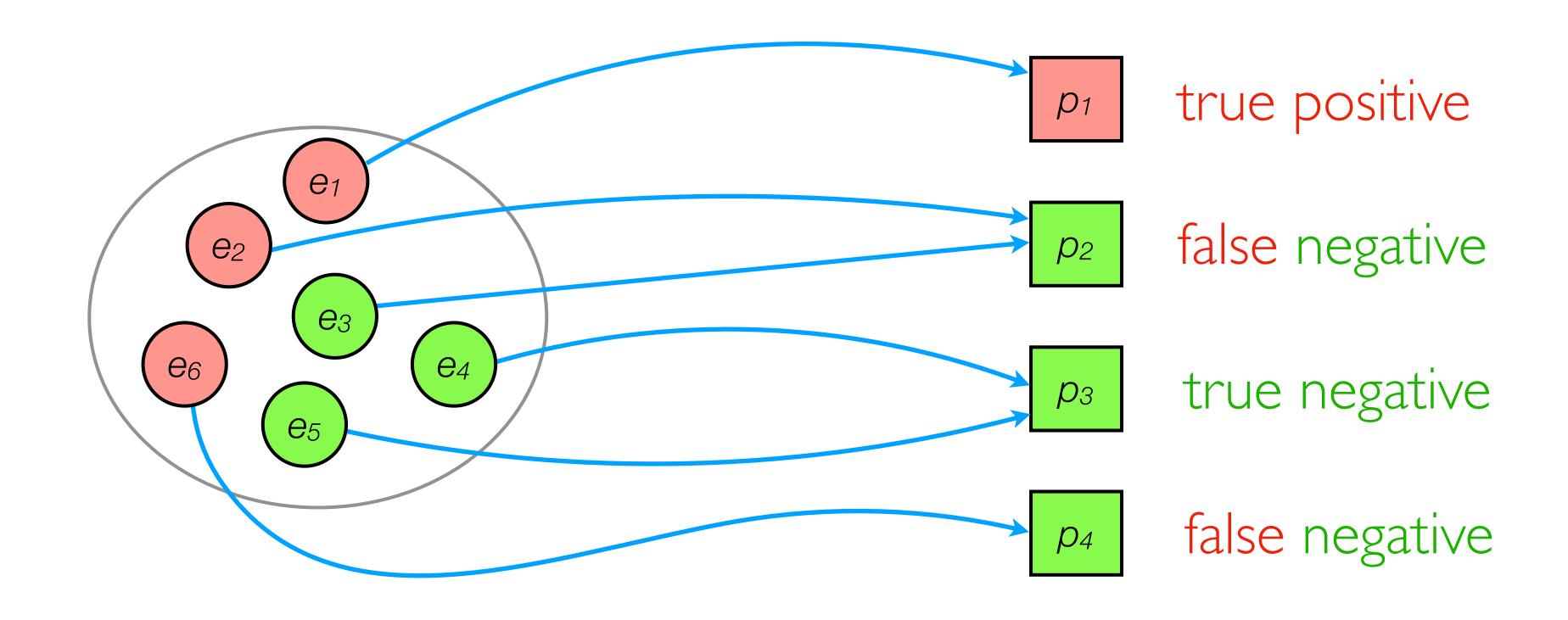
#### Static Bug Finder



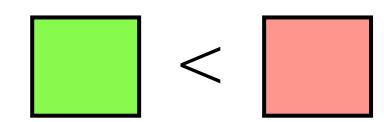
## Unsound Static Bug Finder



#### Sound (but imprecise) Static Bug Finder



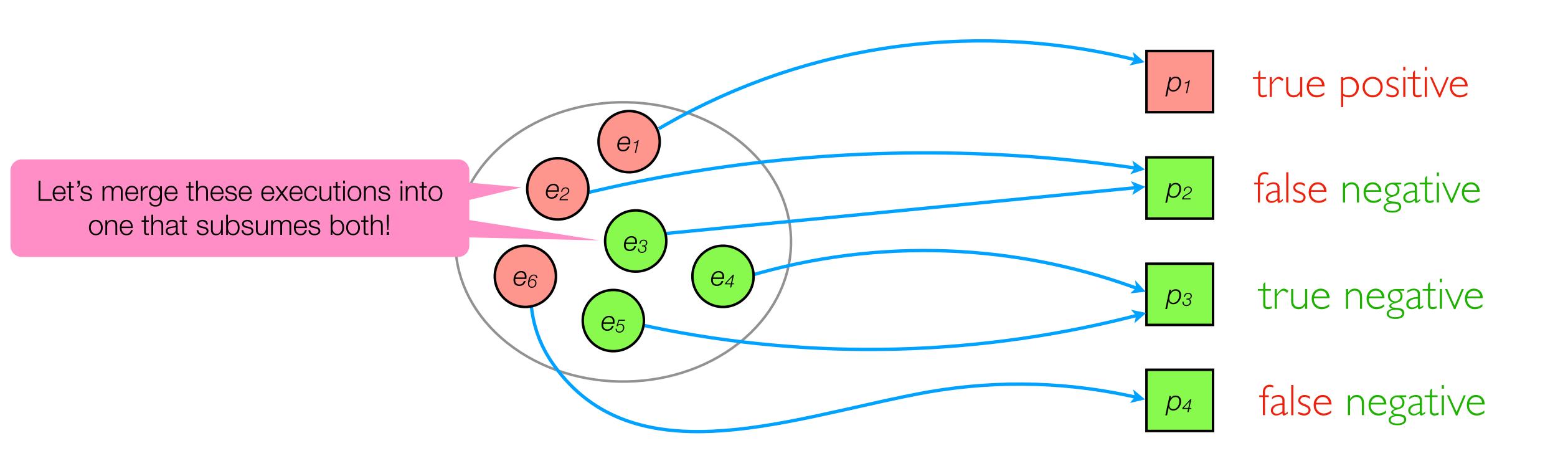
abstract under-approximation

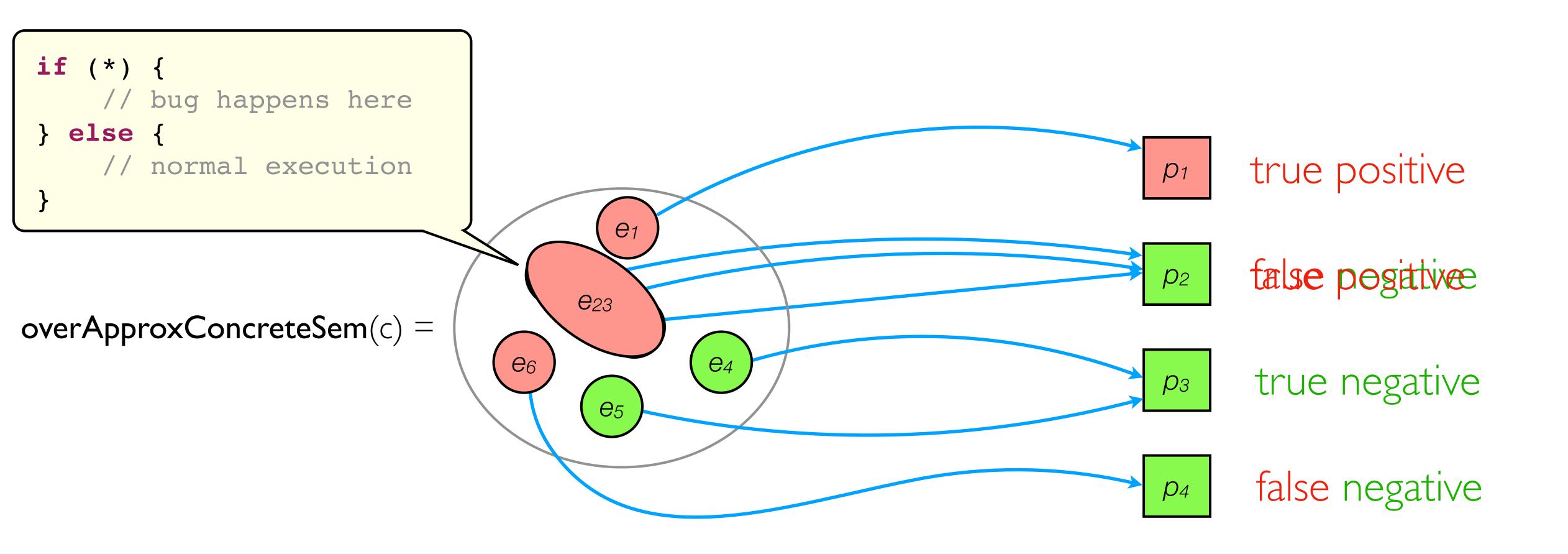


#### Loss of Precision in Static Bug Finders

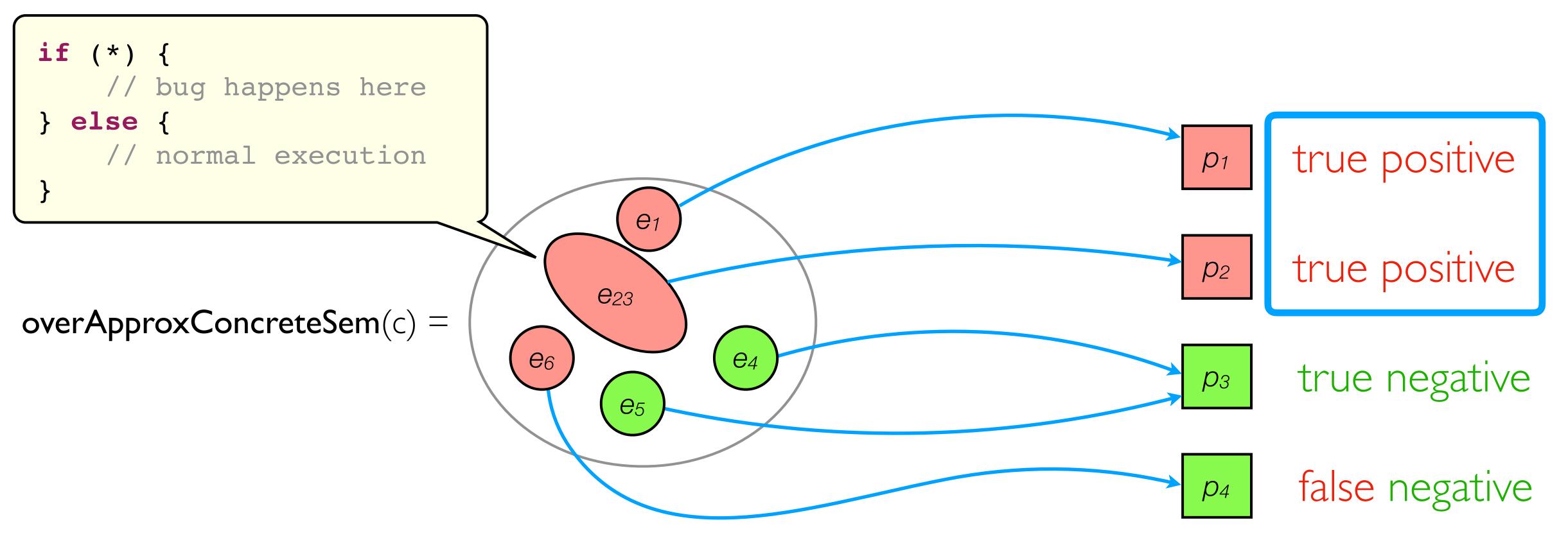
Idea: over-approximate in concrete semantics!

#### Sound (but Imprecise) Static Bug Finder





# Sound Static Bug Finder



concrete over-approximation

abstract under-approximation



#### Towards Sound Static Bug Finders

(this work)

- False negatives (bugs missed) are okay
- False positives (non-bugs reported) are bad
- · Constructed as under-approximation of over-approximation
- Soundness (True Positives) Theorem:
  Under certain assumptions about the programs, the analyser has **no false positives**.

#### A Recipe for True Positives Theorem

- 1. **Over-approximate** semantic elements to make up for "difficult" dynamic execution aspects **Example:** replace conditions and loops with their *non-deterministic* versions
- 2. Pick abstraction  $\alpha$  for over-approximated executions that **provably identifies** "buggy" behaviours:  $\forall$  e: execution, **hasBug**( $\alpha$ (e))  $\Rightarrow$  execution e has a bug
- 3. Design an abstract semantics **asem**, so it is *complete* wrt.  $\alpha$  and over-approximated concrete semantics:  $\forall$  c : program, **asem**(c) =  $\alpha$ (overApproxConcreteSem(c))
- 4. Together, asem and hasBug provide a TP-sound static bug finder.

#### Case Study: RacerDX

- A provably TP-Sound version of Facebook's RacerD concurrency analyser
- Buggy executions: data races in lock-based concurrent programs
- Syntactic assumptions:
  Java programs with well-scoped locking (synchronised), no recursion, reflection, dynamic class loading; global variables are ignored.
- Concrete over-approximation:
   Loops and conditionals are non-deterministic.

#### A True Race

```
class Bloop {
  public int f = 1;
}
```

```
class Burble {
  public void meps(Bloop b) {
    synchronized (this) {
      System.out.println(b.f);
  public void reps(Bloop b) {
   b.f = 42;
  public void beps(Bloop b) {
    b = new Bloop();
    b.f = 239;
```

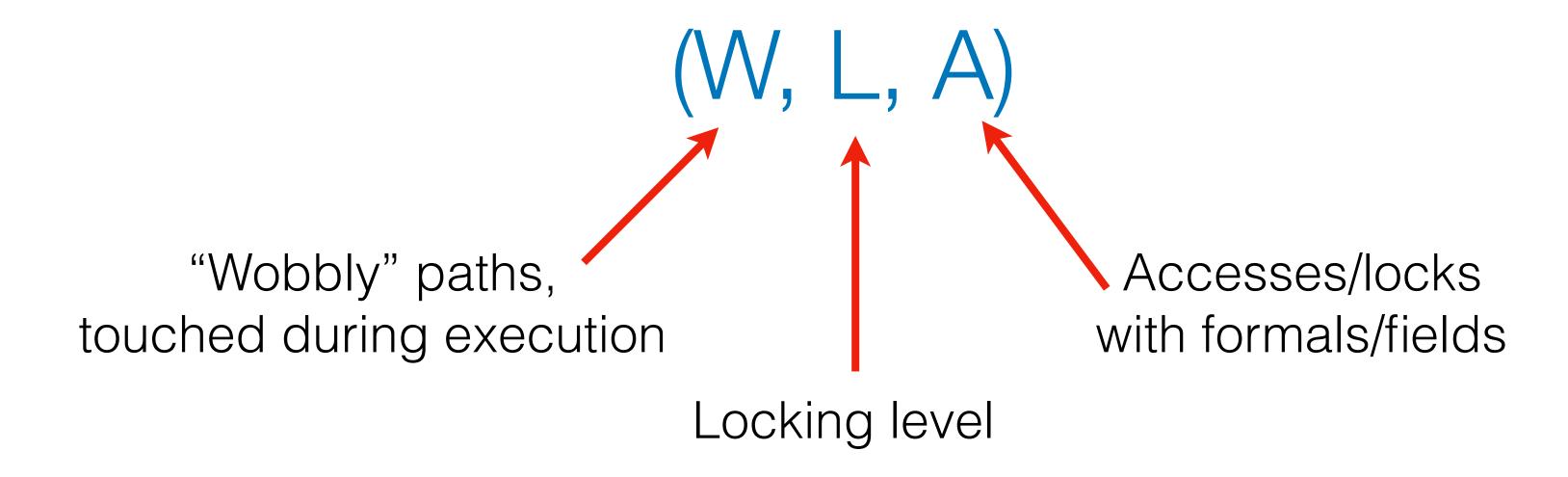
#### A False Race

```
class Bloop {
  public int f = 1;
}
```

Path prefix **b** is "unstable" ("wobbly"), as it's reassigned, hence race is evaded.

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class Burble {
  public void meps(Bloop b) {
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```

#### Complete Abstraction for Race Detection



- $asem(meps(b)) = (\{b,f\}, 0, \{R(b,f, 1)\})$
- $asem(reps(b)) = (\{b,f\}, 0, \{W(b,f, 0)\})$
- $asem(beps(b)) = (\{b, b, b, f\}, 0, \{W(b, 0), W(b, f, 0)\})$

```
class Burble {

public void meps(Bloop b) {
    synchronized (this) {
        System.out.println(b.f);
    }
}

public void reps(Bloop b) {
    b.f = 42;
}

public void beps(Bloop b) {
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```

#### Analysing Summaries for Races

```
• asem(meps(b)) = (\{b,f\}, 0, \{R(b,f, 1)\})
• asem(reps(b)) = (\{b,f\}, 0, \{W(b,f, 0)\})
• asem(beps(b)) = (\{b, b, b, f\}, 0, \{W(b, 0), W(b, f, 0)\})
        meps(b) | reps(b) \Rightarrow Can race,
                                    report a bug!
```

```
class Burble {
 public void meps(Bloop b) {
   synchronized (this) {
     System.out.println(b.f);
  public void reps(Bloop b) {
   b.f = 42;
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#### Analysing Summaries for Races

```
• asem(meps(b)) = (\{b,f\}, 0, \{R(b,f, 1)\})
```

- $asem(reps(b)) = (\{b,f\}, 0, \{W(b,f, 0)\})$
- $asem(beps(b)) = (\{b, b.f\}, 0, \{W(b, 0), W(b.f, 0)\})$

```
meps(b) | beps(b) \Rightarrow Maybe don't race,
```

don't report a bug

```
class Burble {
  public void meps(Bloop b) {
    synchronized (this) {
      System.out.println(b.f);
    }
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public void reps(Bloop b) {
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```

#### Formal Result

# RacerDX enjoys the True Positives Theorem wrt. Data Race Detection

(Details in the paper)

#### Evaluation

What is the price to pay for having the TP Theorem?

(Reporting no bugs whatsoever is TP-Sound)

#### RacerD vs RacerDX

Target	LOC	D CPU	DX CPU	CPU ±%	D Reps	DX Reps	Reps ±%
avrora	76k	103	102	0.4%	143	92	36%
Chronicle-Map	45k	196	196	0.1%	2	2	0%
jvm-tools	33k	106	109	-3.6%	30	26	13%
RxJava	273k	76	69	9.2%	166	134	19%
sunflow	25k	44	44	-1.4%	97	42	57%
xalan-j	175k	144	137	5.0%	326	295	10%

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#### What else is in the POPL'19 paper

- Formal definitions of races, concrete and abstract semantics
- Proof of the TP Theorem for RacerDX
  - Proofs of analysis completeness (wrt. (W, L, A)-abstraction)
  - Proof of the bug detection completeness
- Discussion and comparison with existing static race analyses, e.g., Chord (Naik et al., PLDI'06)

#### To Take Away: Theory

- A True Positive-Sound static bug finder never reports false positives. It can be designed as an under-approximation of an over-approximation
- An abstraction α for TP-Sound static bug detection can be very simple, but it has to be complete (i.e., sufficient) to report bugs.

### To Take Away: Practice

- RacerDX is TP-Sound race detector, whose precision and performance are comparable with Facebook's RacerD
- If RacerDX had been deployed initially rather than RacerD, it would have found 1000s of bugs, far outstripping all *reported impact* in previous concurrency analyses (counterfactual reasoning)
- Until now, static analysers for bug catching that are effective in practice but unsound
  have often been regarded as ad hoc;
   In the future, they can be principled, satisfying theorems to inform and guide their design.