Static Analysis

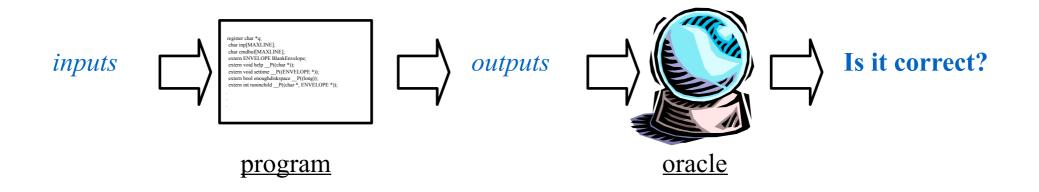
With material from Michelle Mazurek, Dave Levin, Mike Hicks, Dawson Engler, Lujo Bauer, and Jeff Foster



Static Analysis / Symbolic Execution

Current Practice

for Software Assurance



- Testing: Check correctness on set of inputs
- Benefits: Concrete failure proves issue, aids fix
- Drawbacks: Expensive, difficult, coverage?
 - No guarantees

Current Practice

(continued)

- Code audit: Convince someone your code is correct
- Benefit: Humans can generalize
- Drawbacks: Expensive, hard, no guarantees



```
It (streetempte-commander, condown))
break;

**reste errors */
ermo = 0;

*** Process command.

*** If we are running as a null server, return 550

*** to everything.

** to everything.

** if (nullserver)

{
    switch (e>emdcode)

{
    case CMDQUIT:
    case CMDQUIT:
    case CMDPHLO:
    case CMDPHLO:
    case CMDPHLO:
    case CMDPHLO:
    case CMDNOOP:

/* process normally */
break;

default:
    if (++badcommands > MAXBADCOMMANDS)
    sleep(1);
    user(r(*550 Access denied**);
    continue;

}

/* non-null server */
switch (e>emdcode)

{
    case CMDMAIL:
    case CMDDXPN:
    case CMDDXPN:
    case CMDRAFY:
```

If You're Worried about Security...

A malicious adversary is trying to exploit anything you miss!



What more can we do?

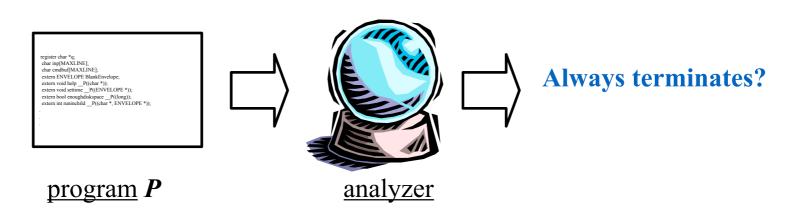
Static analysis

- Analyze program's code without running it
 - In a sense, ask a computer to do code review
- Benefit: (much) higher coverage
 - Reason about many possible runs of the program
 - Sometimes all of them, providing a guarantee
 - Reason about incomplete programs (e.g., libraries)

Drawbacks:

- Can only analyze limited properties
- May miss some errors, or have false alarms
- Can be time- and resource-consuming

The Halting Problem



- Can we write an analyzer that can prove, for any program P and inputs to it, P will terminate?
 - Doing so is called the halting problem
 - Unfortunately, this is undecidable: any analyzer will fail to produce an answer for at least some programs and/or inputs

Check other properties instead?

- Perhaps security-related properties are feasible
 - E.g., that all accesses a[i] are in bounds
- But these properties can be converted into the halting problem by transforming the program
 - A perfect array bounds checker could solve the halting problem, which is impossible!
- Other undecidable properties (Rice's theorem)
 - Does this SQL string come from a tainted source?
 - Is this pointer used after its memory is freed?
 - Do any variables experience data races?

Halting ≈ Index in Bounds

- Change all exits to infinite loops (guaranteed no terminate)
- Change out-of-bounds index to exit:
 - (i >= 0 && i < a.length) ? a[i] : exit()
- Now if the array bounds checker
 - ... finds an error, then the original program halts
 - ... claims there are no such errors, then the original program does not halt
 - ... contradiction! with halting undecidability

So is static analysis impossible?

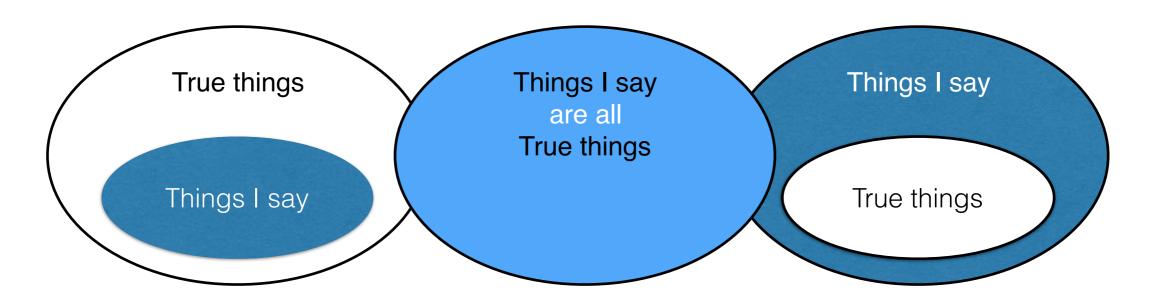
- Perfect static analysis is not possible
- Useful static analysis is perfectly possible, despite
 - 1. Nontermination analyzer never terminates, or
 - 2. False alarms claimed errors are not really errors, or
 - 3. **Missed errors** no error reports ≠ error free
- Nonterminating analyses are confusing, so tools tend to exhibit only false alarms and/or missed errors

Completeness

If analysis says that X is true, then X is true.

Soundness

If X is true, then analysis says X is true.



Trivially Complete: Say nothing Trivially Sound: Say everything

Sound and Complete:
Say exactly the set of true things

Stepping back

- **Soundness**: No error found = no error exists
 - Alarms may be false errors
- Completeness: Any error found = real error
 - Silence does not guarantee no errors
- Basically any useful analysis
 - is neither **sound** nor **complete** (def. not **both**)
 - ... usually *leans* one way or the other
 - Academic analyses lean towards sound

The Art of Static Analysis

- Design goals:
 - Precision: Carefully model program, minimize false positives/negatives
 - Scalability: Successfully analyze large programs
 - Understandability: Error reports should be actionable
- Observation: Code style is important
 - Aim to be precise for "good" programs
 - OK to forbid yucky code in the name of safety
 - Code that is more understandable to the analysis is more understandable to humans

First, a few words on different types of analyses...

Many Kinds of Analyses

Constraint-Based

Shape

Type-Based

- Pointer
- Abstract Interpretation
- Dataflow

Symbolic Execution

Interprocedural

And many, many more!!!

All analyses have one thing in common:

AST

They define how to take each piece of the program and interpret it in some part of the analysis framework

This is what changes!

A few examples...

- Constraints
- Points in Lattice
- Sets of numbers / values

Tainted Flow Analysis

- Cause of many attacks is trusting unvalidated input
 - Input from the user (network, file) is tainted
 - Various data is used, assuming it is untainted
- Examples expecting untainted data
 - source string of strcpy (≤ target buffer size)
 - format string of printf (contains no format specifiers)
 - form field used in constructed SQL query (contains no SQL commands)

Recall: Format String Attack

Adversary-controlled format string

```
char *name = fgets(..., network_fd);
printf(name);  // Oops
```

- Attacker sets name = "%s%s%s" to crash program
- Attacker sets name = "...%n..." to write to memory
 - Yields code injection exploits
- These bugs still occur in the wild occasionally
 - Too restrictive to forbid non-constant format strings

The problem, in types

• Specify our requirement as a type qualifier

```
int printf(untainted char *fmt, ...);
tainted char *fgets(...);
```

- tainted = possibly controlled by adversary
- untainted = must not be controlled by adversary

```
tainted char *name = fgets(...,network_fd);
printf(name); // FAIL: tainted ≠ untainted
```

Analyzing taint flows

- Goal: For all possible inputs, prove tainted data will never be used where untainted data is expected
 - untainted annotation: indicates a trusted sink
 - tainted annotation: an untrusted source
 - no annotation means: not sure (analysis must figure it out)
- Solution requires inferring flows in the program
 - What sources can reach what sinks
 - If any flows are illegal, i.e., whether a tainted source may flow to an untainted sink
- We will aim to develop a sound analysis

Legal Flow

```
void f(tainted int);
untainted int a = ...;
f(a);
```

f accepts tainted or untainted data

untainted

tainted

Illegal Flow

```
void g(untainted int);
tainted int b = ...;
g(b);
```

g accepts only untainted data

tainted \(\preceq \) untainted

Define allowed flow as a **lattice**:

untainted < tainted

At each program step, **test** whether inputs ≤ policy

Analysis Approach

- If no qualifier is present, we must **infer** it
- Steps:
 - Create a name for each missing qualifier (e.g., α, β)
 - For each program statement, generate constraints
 - Statement x = y generates constraint $q_y \le q_x$
 - Solve the constraints to produce solutions for α , β , etc.
 - A solution is a substitution of qualifiers (like tainted or untainted) for names (like α and β) such that all of the constraints are legal flows
- If there is no solution, we (may) have an illegal flow

Example Analysis

```
int printf(untainted char *fmt, ...);
tainted char *fgets(...);

a char *name = fgets(..., network_fd);
β char *x = name;
printf(x);
```

- 1 tainted ≤ α
- \bigcirc $\alpha \leq \beta$
- $\beta \leq untainted$

Illegal flow!

First constraint requires α = tainted To satisfy the second constraint implies β = tainted But then the third constraint is illegal: tainted \leq untainted

Taint Analysis: Adding Sensitivity



But what about?

```
int printf(untainted char *fmt, ...);
tainted char *fgets(...);

\( \alpha \text{ char *name = fgets(..., network_fd);} \)
\( \beta \text{ char *x;} \)
\( \text{x = name;} \)
\( \text{x = "hello!";} \)
\( \text{printf(x);} \)

tainted \( \leq \alpha \)
\( \alpha \leq \beta \)
\( \text{untainted} \leq \beta \)
\( \text{No constraint solution. Bug?} \)
\( \beta \leq \text{untainted} \)
\( \beta \leq \text{untainted} \)
\( \text{False Alarm!} \)
\( \text{Volume of the print for the p
```

Flow Sensitivity

- Our analysis is flow insensitive
 - Each variable has one qualifier
 - Conflates the taintedness of all values it ever contains
- Flow-sensitive analysis accounts for variables whose contents change
 - Allow each assigned use of a variable to have a different qualifier
 - E.g., α₁ is x's qualifier at line 1, but α₂ is the qualifier at line 2, where α₁ and α₂ can differ
 - Could implement this by transforming the program to assign to a variable at most once

Reworked Example

```
int printf(untainted char *fmt, ...);
tainted char *fgets(...);
```

```
tainted \leq \alpha

\alpha \leq \beta

untainted \leq \gamma

\gamma \leq untainted
```

No Alarm

Good solution exists:

y = untainted

 $\alpha = \beta = tainted$

Handling conditionals

```
int printf(untainted char *fmt, ...);
tainted char *fgets(...);

α char *name = fgets(..., network_fd);
```

```
β char *x;

if (...) x = name;

else x = "hello!";

printf(x);
```

```
tainted \leq \alpha

\alpha \leq \beta

untainted \leq \beta

\beta \leq untainted
```

Constraints still unsolvable **Illegal flow**

Multiple Conditionals

```
int printf(untainted char *fmt, ...);
tainted char *fgets(...);
```

```
void f(int x) {
    α char *y;
    if (x) y = "hello!";
    else y = fgets(..., network_fd);
    if (x) printf(y);
}
```

-untainted ≤ α

tainted $\leq \alpha$

 $\alpha \leq untainted$

No solution for α . Bug?

False Alarm!

(and flow sensitivity won't help)

Path Sensitivity

- Consider path feasibility. E.g., f(x) can execute path
 - 1-2-4-5-6 when $x \neq 0$, or
 - 1-3-4-6 when x == 0. But,
 - path 1-3-4-5-6 infeasible

```
void f(int x) {
  char *y;
  lf (x) 2y = "hello!";
  else 3y = fgets(...);
  lf (x) printf(y);
6}
```

 A path sensitive analysis checks feasibility, e.g., by qualifying each constraint with a path condition

```
    x ≠ 0 ⇒ untainted ≤ α (segment 1-2)
    x = 0 ⇒ tainted ≤ α (segment 1-3)
    x ≠ 0 ⇒ α ≤ untainted (segment 4-5)
```

Why not use flow/path sensitivity?

- Flow sensitivity adds precision, path sensitivity adds more
 - Reduce false positives: less developer effort!
- But both of these make solving more difficult
 - Flow sensitivity increases the number of nodes in the constraint graph
 - Path sensitivity requires more general solving procedures to handle path conditions
- In short: precision (often) trades off scalability
 - Ultimately, limits the size of programs we can analyze

Handling Function Calls

```
α char *a = fgets(...);
β char *b = id(a);
```

```
δ char *id(Y char *x) {
  return x;
}
```

- Names for arguments and return value
- Calls create flows
 - from caller's data to callee's arguments,
 - from callee's result to caller's returned value

Handling Function Calls

```
α char *a = fgets(...);
β char *b = id(a);

tainted < α
```

```
\begin{array}{c} \text{tainted} \leq \alpha \\ \alpha \leq \gamma \\ \gamma \leq \delta \\ \delta \leq \beta \end{array}
```

Result: b is tainted (as expected)

Function Call Example

```
a char *a = fgets(...);
β char *b = id(a);
w char *c = "hi";
printf(c);
```

```
δ char *id(Y char *x) {
  return x;
}
```

```
\begin{array}{c} \text{tainted} \leq \alpha \\ \alpha \leq \gamma \\ \gamma \leq \delta \\ \delta \leq \beta \\ \text{untainted} \leq \omega \end{array} \qquad \begin{array}{c} \text{No Alarm} \\ \text{Good solution exists:} \\ \omega = \text{untainted} \\ \alpha = \beta = \gamma = \delta = \text{tainted} \\ \text{untainted} \leq \omega \\ \end{array}
```

ω ≤ untainted

Two Calls to Same Function

```
α char *a = fgets(...);
β char *b = id(a);
ω char *c = id("hi");
printf(c);
```

```
δ char *id(Y char *x) {
  return x;
}
```

```
\begin{array}{l} \text{tainted} \leq \alpha \\ \alpha \leq \gamma \\ \gamma \leq \delta \\ \delta \leq \beta \\ \text{untainted} \leq \gamma \\ \delta \leq \omega \\ \omega \leq \text{untainted} \end{array}
```

No solution. Real bug?

False Alarm!

Two Calls to Same Function

```
α char *a = fgets(...);
β char *b = id(a);
ω char *c = id("hi");
printf(c);

δ char *id(Y char *x) {
 return x;
}
```

tainted $\leq \alpha \leq \gamma \leq \delta \leq \omega \leq untainted$

Problematic constraints represent an infeasible path

False Alarm!

Context (In)sensitivity

- This is a problem of context insensitivity
 - All call sites are "conflated" in the graph
- Context sensitivity solves this problem by:
 - Labeling call sites in some way (e.g. line number)
 - Matching calls with the corresponding returns
 - Label call and return edges
 - Allow flows if the labels match

Two Calls to Same Function

```
α char *a = fgets(...); δ char *id(γ char *x) {
β char *b = id<sub>1</sub>(a); return x;
γ char *c = id<sub>2</sub>("hi");
ρ rintf(c);
```

```
tainted \leq \alpha
0 \leq 1 \text{ y}
\gamma \leq \delta
\frac{\delta \leq 1 \text{ }\beta}{\delta \leq 2 \text{ y}}
\omega \leq \text{untainted}
```

Indexes don't match up

Infeasible flow not allowed

No Alarm

Discussion

- Context sensitivity: another precision/scalability tradeoff
 - O(n) insensitive algorithm becomes $O(n^3)$ sensitive algorithm
 - But: Eliminates infeasible paths (makes n smaller)
 - Sometimes higher precision improves performance
- Compromises possible
 - Only some call sites treated sensitively
 - Conflate groups of call sites
 - Sensitivity only up to a certain call depth

Flow Analysis: Scaling it up to a complete language and problem set



Pointers

```
α char *a = "hi";
(β char *)*p = &a;
(γ char *)*q = p;
ω char *b = fgets(...);
*q = b;
printf(*p);
```

```
Solution exists:
```

```
\alpha = \beta = untainted
\omega = \gamma = tainted
```

Misses illegal flow!

```
\begin{array}{c} \text{untainted} \leq \alpha \\ & \alpha \leq \beta \\ & \beta \leq \gamma \\ & \text{tainted} \leq \omega \\ & \omega \leq \gamma \\ & \beta \leq \text{untainted} \end{array}
```

```
    p and q are aliases
    -so writing tainted data to q
    -makes p's contents
```

tainted

Pointers

```
α char *a = "hi";

(β char *)*p = &a;

(γ char *)*q = p;

ω char *b = fgets(...);

*q = b;

printf(*p);
```

```
Solution exists:

\alpha = \beta = \text{untainted}

\omega = \gamma = \text{tainted}
```

```
\begin{array}{c} \text{untainted} \leq \alpha \\ & \alpha \leq \beta \\ & \beta \leq \gamma \\ & \text{taintev} \leq \beta \nu \\ & \omega \leq \gamma \\ & \beta \leq \text{untainted} \end{array}
```

Pointers

```
α char *a = "hi";

(β char *)*p = &a;

(γ char *)*q = p;

ω char *b = fgets(...);

*q = b;

printf(*p);
```

```
Solution exists:

\alpha = \beta = \text{untainted}

\omega = \gamma = \text{tainted}
```

```
\begin{array}{l} \text{untainted} \leq \alpha \\ \alpha \leq \beta \\ \beta \leq \gamma \\ \gamma \leq \beta \\ \text{tainted} \leq \omega \\ \omega \leq \gamma \\ \beta \leq \text{untainted} \end{array}
```

Flow and pointers

- An assignment via a pointer "flows both ways"
 - Ensures that aliasing constraints are sound
 - But can lead to false alarms
- Reducing alarms
 - If pointers are never assigned to (const)
 then backward flow is not needed (sound)
 - Drop backward flow edge anyway
 - Trades false alarms for missed errors (unsoundness)

Implicit flows

Illegal flow: tainted ≰ untainted

Implicit flows

Missed flow!

Implicit flow analysis

- Implicit flow: one value implicitly influences another
- One way to find these: maintain a scoped program counter (pc) label
 - Represents the maximum taint affecting the current pc
- Assignments generate constraints involving the pc
 - x = y produces two constraints:

```
label(y) \le label(x) (as usual)

pc \le label(x)
```

Implicit flow example

```
tainted int src;

    int dst;

                   if (src == 0)
pc_1 = untainted
pc_2 = tainted
                     dst = 0;
                                                untainted < α
                                                pc_2 \leq \alpha
                   else
pc_3 = tainted
                    dst = 1;
                                                untainted \leq \alpha
                                                DC3 \leq \mathbf{C}
pc_4 = untainted | dst += 0;
                                                untainted \leq \alpha
                                                DC_4 \leq \mathbf{C}
```

: tainted ≤ α

Taint on α is identified.

Discovers implicit flow!

Why not implicit flow?

- Tracking implicit flows can lead to false alarms
 - E.g., ignores values

- Extra constraints hurt performance
- The evil copying example is pathological
 - We typically don't write programs like this*
 - Implicit flows will have little overall influence
- So: taint analyses tend to ignore implicit flows

Other challenges

- Taint through operations
 - tainted a; untainted b; c=a+b is c tainted? (yes, probably)
- Function calls and context sensitivity
 - Function pointers: Flow analysis to compute possible targets
- Struct fields
 - Track taint for the whole struct, or each field?
 - Taint per instance, or shared among all of them (or something in between)?
 - Note: objects ≈ structs + function pointers
- Arrays: Track taint per element or across whole array?

No single correct answer!

(Tradeoffs: Soundness, completeness, performance)

Other refinements

- Label additional sources and sinks
 - e.g., Array accesses must have untainted index
- Handle sanitizer functions
 - Convert tainted data to untainted
- Complementary goal: Leaking confidential data
 - Don't want secret sources to go to public sinks
 - Implicit flows more relevant (malicious code)
 - Dual of tainting

Other kinds of analysis

- Pointer Analysis ("points-to" analysis)
 - Determine whether pointers point to the same locations
 - Shares many elements of flow analysis. Really advanced in the last 10 years.

Data Flow Analysis

 Invented in the early 1970's. Flow sensitive, tracks "data flow facts" about variables in the program

Abstract interpretation

- Invented in the late 1970's as a theoretical foundation for data flow analysis, and static analysis generally.
- Associated with certain analysis algorithms

Symbolic Execution

Introduction

- Static analysis is great
 - Lots of interesting ideas and tools
 - Commercial companies sell, use static analysis
 - It all looks good on paper, and in papers

- But can developers use it?
 - Our experience: Not easily
 - Results in papers describe use by static analysis experts
 - Commercial tools have a huge code mass to deal with developer confusion, false positives, warning management, etc 54

One Issue: Abstraction

- Abstraction lets us scale and model all possible runs
 - But it also introduces conservatism
 - *-sensitivities attempt to deal with this
 - * = flow-, context-, path-, field-, etc
 - But they are never enough

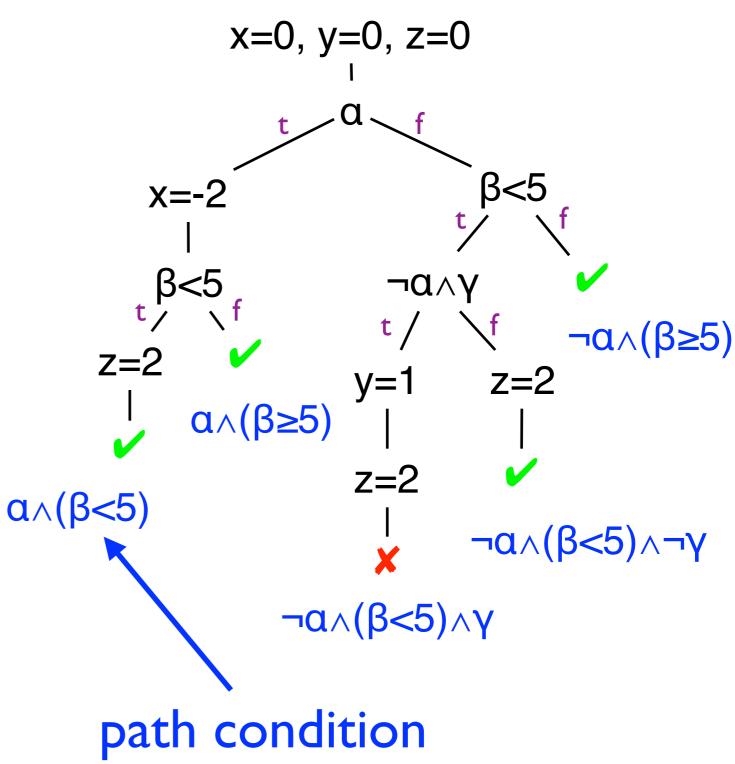
- Static analysis abstraction # developer abstraction
 - Because the developer didn't have them in mind

Symbolic Execution

- Testing works
 - But, each test only explores one possible execution
 - $\operatorname{assert}(f(3) == 5)$
 - We hope test cases generalize, but no guarantees
- Symbolic execution generalizes testing
 - Allows unknown symbolic variables in evaluation
 - $y = \alpha$; assert(f(y) == 2*y-1);
 - If execution path depends on unknown, conceptually fork symbolic executor
 - int f(int x) { if (x > 0) then return 2*x 1; else return 10; }

Symbolic Execution Example

```
1. int a = \alpha, b = \beta, c = \gamma;
   // symbolic
3. int x = 0, y = 0, z = 0;
4. if (a) {
5. x = -2;
7. if (b < 5) {
8. if (!a \&\& c) \{ y = 1; \}
9. z = 2;
10. }
11. assert(x+y+z!=3)
```



Insight

- Each symbolic execution path stands for many actually program runs
 - In fact, exactly the set of runs whose concrete values satisfy the path condition
- Thus, we can cover a lot more of the program's execution space than testing can

Early work on symbolic execution

- Robert S. Boyer, Bernard Elspas, and Karl N. Levitt.
 SELECT—a formal system for testing and debugging programs by symbolic execution. In ICRS, pages 234–245, 1975.
- James C. King. Symbolic execution and program testing. CACM, 19(7):385–394, 1976. (most cited)
- Leon J. Osterweil and Lloyd D. Fosdick. Program testing techniques using simulated execution. In ANSS, pages 171–177, 1976.
- William E. Howden. Symbolic testing and the DISSECT symbolic evaluation system. IEEE Transactions on Software Engineering, 3(4):266–278, 1977.

The problem

- Computers were small (not much memory) and slow (not much processing power) then
 - Apple's iPad 2 is as fast as a Cray-2 from the 1980's

- Symbolic execution is potentially extremely expensive
 - Lots of possible program paths
 - Need to query solver a lot to decide which paths are feasible, which assertions could be false
 - Program state has many bits

Today

- Computers are much faster, memory is cheap
- There are very powerful SMT/SAT solvers today
 - SMT = Satisfiability Modulo Theories = SAT++
 - Can solve very large instances, very quickly
 - Lets us check assertions, prune infeasible paths
 - We've used Z3, STP, and Yices
- Recent success: bug finding
 - Heuristic search through space of possible executions
 - Find really interesting bugs

Path explosion

- Usually can't run symbolic execution to exhaustion
 - Exponential in branching structure

```
    int a = α, b = β, c = γ; // symbolic
    if (a) ... else ...;
    if (b) ... else ...;
    if (c) ... else ...;
```

- Ex: 3 variables, 8 program paths
- Loops on symbolic variables even worse

```
    int a = α; // symbolic
    while (a) do ...;
    3.
```

Potentially 2^31 paths through loop!

Search strategies

- Need to prioritize search
 - Try to steer search towards paths more likely to contain assertion failures
 - Only run for a certain length of time
 - So if we don't find a bug/vulnerability within time budget, too bad
- Think of program execution as a dag
 - Nodes = program states
 - Edge(n1,n2) = can transition from state n1 to state n2
- Then we need some kind of graph exploration strategy
 - At each step, pick among all possible paths

Basic search

- Simplest ideas: algorithms 101
 - Depth-first search (DFS)
 - Breadth-first search (BFS)
 - Which of these did we implement?
- Potential drawbacks
 - Neither is guided by any higher-level knowledge
 - Probably a bad sign
 - DFS could easily get stuck in one part of the program
 - E.g., it could keep going around a loop over and over again
 - Of these two, BFS is a better choice

Randomness

- We don't know a priori which paths to take, so adding some randomness seems like a good idea
 - Idea 1: pick next path to explore uniformly at random (Random Path, RP)
 - Idea 2: randomly restart search if haven't hit anything interesting in a while
 - Idea 3: when have equal priority paths to explore, choose next one at random
 - All of these are good ideas, and randomness is very effective
- One drawback: reproducibility
 - Probably good to use psuedo-randomness based on seed, and then record which seed is picked
 - (More important for symbolic execution implementers than users)

Coverage-guided heuristics

- Idea: Try to visit statements we haven't seen before
- Approach
 - Score of statement = # times it's been seen and how often
 - Pick next statement to explore that has lowest score
- Why might this work?
 - Errors are often in hard-to-reach parts of the program
 - This strategy tries to reach everywhere.
- Why might this not work?
 - Maybe never be able to get to a statement if proper precondition not set up
- KLEE = RP + coverage-guided

Generational search

- Hybrid of BFS and coverage-guided
- Generation 0: pick one program at random, run to completion
- Generation 1: take paths from gen 0, negate one branch condition on a path to yield a new path prefix, find a solution for that path prefix, and then take the resulting path
 - Note will semi-randomly assign to any variables not constrained by the path prefix
- Generation n: similar, but branching off gen n-1
- Also uses a coverage heuristic to pick priority

Combined search

- Run multiple searches at the same time
- Alternate between them
 - E.g., Fitnext
- Idea: no one-size-fits-all solution
 - Depends on conditions needed to exhibit bug
 - So will be as good as "best" solution, which a constant factor for wasting time with other algorithms
 - Could potentially use different algorithms to reach different parts of the program

SMT solver performance

- SAT solvers are at core of SMT solvers
 - In theory, could reduce all SMT queries to SAT queries
 - In practice, SMT and higher-level optimizations are critical

Some examples

- Simple identities (x + 0 = x, x * 0 = 0)
- Theory of arrays (read(42, write(42, x, A)) = x)
 - 42 = array index, A = array, x = element
- Caching (memoize solver queries)
- Remove useless variables
 - E.g., if trying to show path feasible, only the part of the path condition related to variables in guard are important

Libraries and native code

- At some point, symbolic execution will reach the "edges" of the application
 - Library, system, or assembly code calls
- In some cases, could pull in that code also
 - E.g., pull in libc and symbolically execute it
 - But glibc is really complicated
 - Symbolic execution can easily get stuck in it
 - ⇒ pull in a simpler version of libc, e.g., newlib
 - libc versions for embedded systems tend to be simpler
- In other cases, need to make models of code
 - E.g., implement ramdisk to model kernel fs code
 - This is a lot of work!

Concolic execution

- Also called dynamic symbolic execution
- Instrument the program to do symbolic execution as the program runs
 - I.e., shadow concrete program state with symbolic variables
- Explore one path, from start to completion, at a time
 - Thus, always have a concrete underlying value to rely on

Concretization

- Concolic execution makes it really easy to concretize
 - Replace symbolic variables with concrete values that satisfy the path condition
 - Always have these around in concolic execution
- So, could actually do system calls
 - But we lose symbolic-ness at such calls
- And can handle cases when conditions too complex for SMT solver
 - But can do the same in pure symbolic system

Resurgence of symbolic exection

- Two key systems that triggered revival of this topic:
 - DART Godefroid and Sen, PLDI 2005
 - Godefroid = model checking, formal systems background
 - EXE Cadar, Ganesh, Pawlowski, Dill, and Engler, CCS 2006
 - Ganesh and Dill = SMT solver called "STP" (used in implementation)
 - Theory of arrays
 - Cadar and Engler = systems

KLEE: Coreutils crashes

```
paste -d\\ abcdefghijklmnopqrstuvwxyz
pr -e t2.txt
tac -r t3.txt t3.txt
mkdir -Z a b
mkfifo -Z a b
mknod -Z a b p
md5sum -c t1.txt
ptx -F\\ abcdefghijklmnopqrstuvwxyz
ptx x t4.txt
seq -f %0 1
t1.txt: "\t \tMD5("
t2.txt: "\b\b\b\b\b\b\b\t"
t3.txt: "\n"
t4.txt: "a"
```

Figure 7: KLEE-generated command lines and inputs (modified for readability) that cause program crashes in COREUTILS version 6.10 when run on Fedora Core 7 with SELinux on a Pentium machine.

Static analysis in practice

- Thoroughly check limited but useful properties
 - Eliminate some categories of errors
 - Developers can concentrate on deeper reasoning
- Encourage better development practices
 - Programming models that avoid mistakes
 - Teach programmers to manifest their assumptions
 - Using annotations that improve tool precision
- Seeing increased commercial adoption