



EDAP15: Program Analysis

DYNAMIC PROGRAM ANALYSIS 2

Christoph Reichenbach

- Interrupts program during execution
- Examines call stack

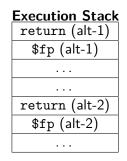
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 - ► Simulates CPU/Memory in software
 - Tries to replicate inner workings of machine
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 - Counts important system events (network accesses etc.)
- ► CPU:
 - Hardware performance counters count interesting events

Profiler

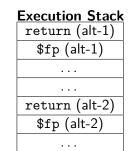
Measures: which functions are we spending our time in?



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- ► Approach:
 - Build stack maps
 - Execute program, interrupt regularly
 - During interrupt:
 - Examine stack
- Infer functions from stack contents

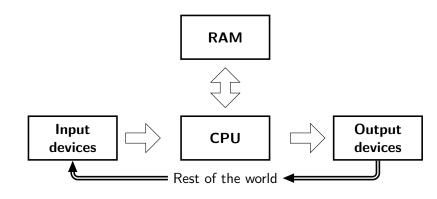
Execution Stack return (alt-1)
\$fp (alt-1)
return (alt-2)
\$fp (alt-2)

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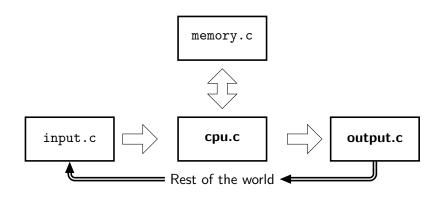


Can be inaccurate: misses short function calls

Simulator

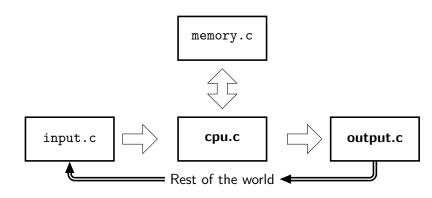


Simulator



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- Can count events of interest (memory accesses etc.)

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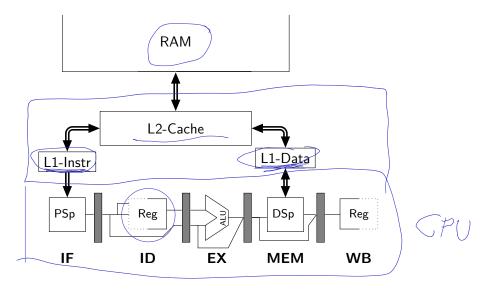
Modern CPUs are very complex: Simulators tend to be inaccurate

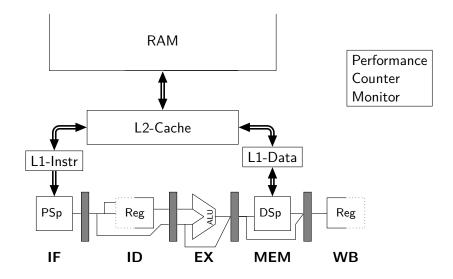
Software Performance Counters

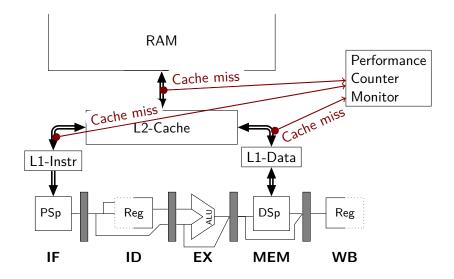
- Complex software may use high-level properties such as:
 - How much time do we spend waiting for the harddisk?
 - How often was our program suspended by the operating system in order to let another program run?
 - How much data did we receive through the network?

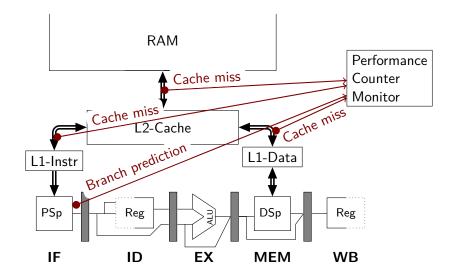
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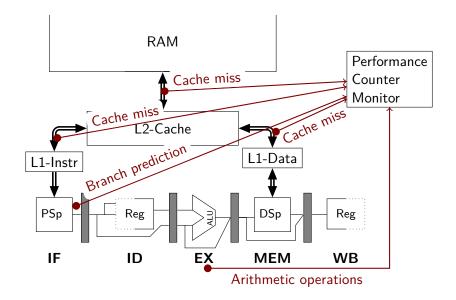
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 - How much time do we spend waiting for the harddisk?
 - How often was our program suspended by the operating system in order to let another program run?
 - How much data did we receive through the network?
- Operating systems collect many of these statistics











Special CPU registers:

. . .

- Count performance events
- Registers must be configured to collect specific performance events
 - Number of CPU cycles
 - Number of instructions executed
 - Number of memory accesses

► #performance event types > #performance registers

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- ▶ #performance event types > #performance registers

May be inaccurate: not originally built for software developers

Summary

- ▶ Performance analysis may require detailed dynamic data
- **Profiler**: Probes stack contents at certain intervals
- Simulator:
 - Simulates hardware in software, measures
 - Tends to be inaccurate

Performance Counters:

- Software:
 - Operating System counts events of interest
- Hardware:
 - ▶ Special registers can be configured to measure CPU-level events

Measured performance properties are valid for...

- Selected CPU
- Selected operating system

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- CPU configuration (CPU frequency etc.)

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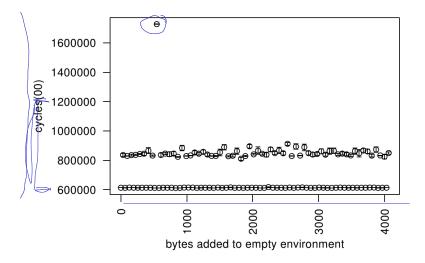
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Changing your user name can speed up code



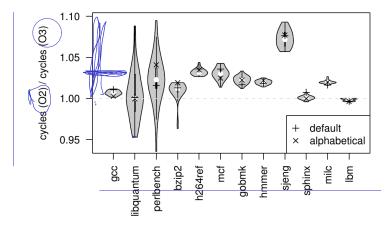
Mytkowicz, Diwan, Hauswirth, Sweeney: "Producing wrong data without doing anything obviously wrong", in ASPLOS 2009

Linking Order

Is there a difference between re-ordering modules in RAM?
gcc a.o b.o -o program (Variant 1)
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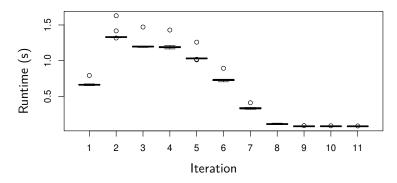
(Mytkowicz, Diwan, Hauswirth, Sweeney, ASPLOS'09)

Adaptive Systems

Measurement: 11 runs

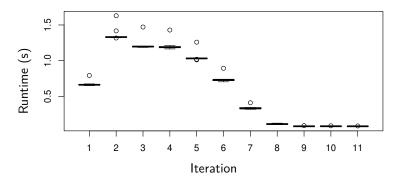
Adaptive Systems





Adaptive Systems





Warm-up effect

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- Reason: Adaptive Systems
 - Hardware:
 - Cache: Speed up some memory accesses
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 - Translation Lookaside Buffer
 - Software:
 - Operating System / Page Table
 - Operating System / Scheduler
 - Just-in-Time compiler

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- What sbould we measure?
 - Latency: measure first run Reset system before every run
 - Throughput: later runs
 Discard initial n measurements

Ignored Parameters

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Beware of generalising measurement results!

Summary

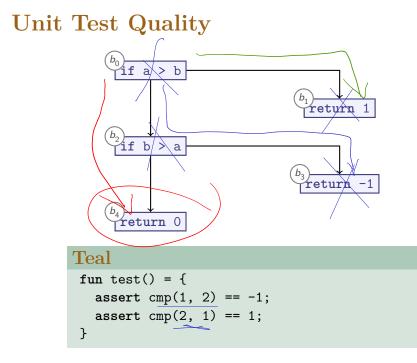
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- Modern computers are complex
 - Caches make memory access times hard to predict
 - Multi-tasking may cause sudden interruptions
- This makes measurements difficult:
 - Must carefully consider what assumptions we are making
 - Must measure repeatedly to gather distribution
 - Must check for warm-up effects
 - Must try to understand causes for performance changes
- Measurements are often not normally distributed
 - ▶ Mean + Standard Deviation may not describe samples well
 - If in doubt, use box plots or violin plots

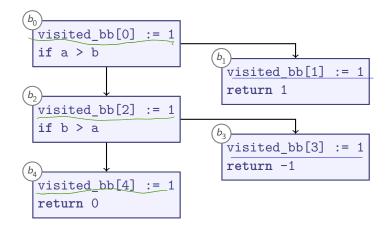
Unit Tests

Teal

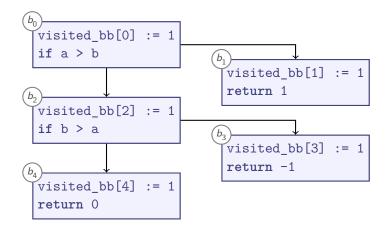
```
fun cmp(a, b) = {
  if a > b \{
    return 1;
  }
  if a < b {
   return -1;
  }
  return 0;
}
fun test() = {
  assert cmp(1, 2) == -1;
  assert cmp(2, 1) == 1;
}
```



Test Coverage



Test Coverage



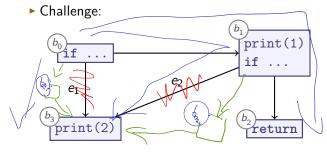
> Test coverage = fraction of visited_bb elements updated

Statement Coverage: is each statement executed?

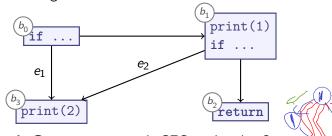
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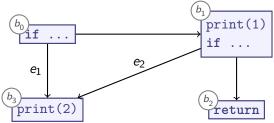


- ► Statement Coverage: is each statement executed?
 ⇔ each Basic Block is executed
- Edge Coverage: is each CFG edge taken?
 - Challenge:



Path Coverage: is each CFG path taken?

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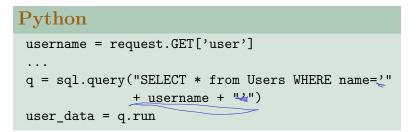
Path Coverage: is each CFG path taken?

- Need to limit Number of loop iterations checked
- Must restart tracking block coverage on every method entry

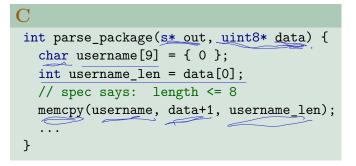
Summary

• Unit Tests are a simple form of dynamic program analysis

- Minimal tooling needed
- Custom checks
- Limited to what underlying language can express directly
- Test Coverage tells us how much of our code gets analysed by at least one unit test
- Implement by setting markers on relevant basic blocks
- Different criteria, such as:
 - Statement Coverage
 - Edge Coverage: may require helper BBs
 - > Path Coverage: paths through CFG (usually excluding loops)



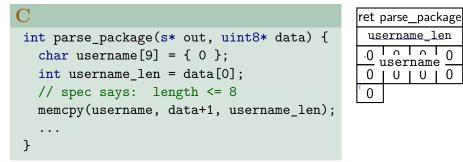




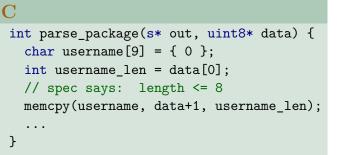
Stack

username

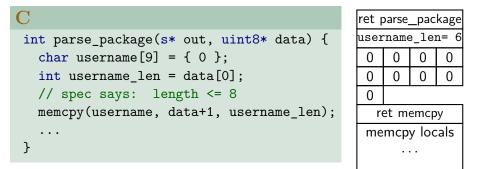
() 1 ()

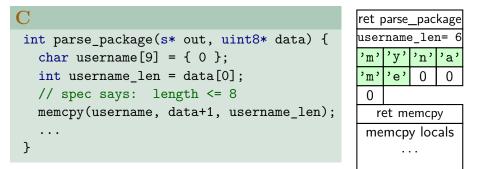


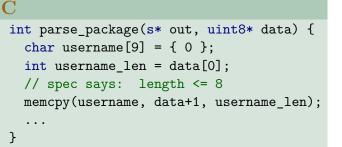
Stack

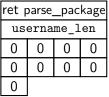


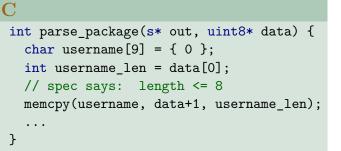
ret parse_package username_len= 6 0 0 0 0 0 0 0 0 0

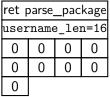


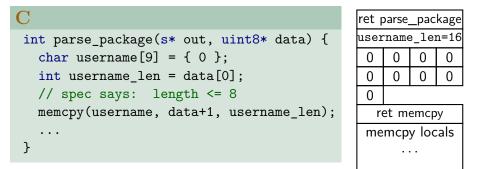


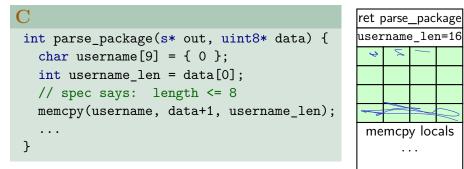












Tracing 'Tainted' Values

Taint Analysis:

- Track tainted values
- Remove taint if values are sanitised
- Detect if they reach sensitive sinks
- ► NB: Static taint analysis may also be possible

Unsafe input

- **Taint source**: Network ops
- Sanitiser: SQL string escape
- Taint sink: SQL query string

Leaking secrets

- Taint source: Plaintext passwd.
- Sanitiser: cryptographic hash
- Taint sink: Network ops

```
~ query_l = "SELECT ...'"
		 query_r = "'"
		 username = request.GET['user']
		 ...
		 query_str = query_l + username
		 query_str = query_str + query_r
		 q = sql.query(query_str)
```

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query_l = "SELECT \dots "<sup>\epsilon</sup>
query_r = "''
```

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```
query_l = "SELECT ...'"
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. . .
query_str = query_1 + username query_str = "..."<sup>t</sup>
query_str = query_str + query_r
q = sql.query(query_str)
```

query_l = "SELECT \dots " ϵ query_r = "''

Dynamic Taint Analysis

Strategy:

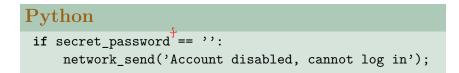
Annotate tainted values with taint tags or shadow values

Check taint sinks for tainted input

Needs instrumentation (shadow values) or explicit support by runtime (e.g., Perl, Ruby)

Conditionals

Should conditionals propagate taint?



Conditionals

- Should conditionals propagate taint?
- Usually such control dependencies don't propagate taint

Python

if secret_password == '':
 network_send('Account disabled, cannot log in');

Attackers vs. Taint Ananlysis

Is taint analysis 'sound enough' to detect attempts to expose sensitive data?

- Often-proposed technique: Taint analysis in Dalvik VM
- Can attackers subvert this analysis?

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```
C
if (secret_password[i] & 1) {
    network_send("Meaninless Message");
} else {
    network_send("Something Else");
}
```

Attackers vs. Taint Ananlysis

Is taint analysis 'sound enough' to detect attempts to expose sensitive data?

- ▶ Often-proposed technique: Taint analysis in Dalvik VM
- Can attackers subvert this analysis?

```
C
for (i = 0; i < 16; ++i) {
  for (k = 0; k < 8; ++k) {
    if (secret_password[i] & 1 << k) {
      network_send("Meaninless Message");
    } else {
      network_send("Something Else");
    }
}</pre>
```

System Command Attack

```
C
 char d secret[1024];
 strcpy(d_secret, "/tmp/");
 strcat(d secret, secret); // taint d secret
 int iopipes[2];
 pipe(iopipes);
 if (fork()) { // create child process
   // connect pipes
   execv("/bin/rm", d_secret); // call external 'rm'
 }
 char[1024] buf; // untained!
 read(iopipes[0], ...); // read output from 'rm'
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System call will print e.g.: rm: cannot remove '/tmp/mysecretstring': No such file or directory

Side Channel Attacks

Many more attacks possible:

- Timing attacks:
 - Two threads
 - One sends signal to other, with delays
 - Delay loop length dependent on secret
- File length attack:
 - Write dummy file
 - ▶ File length (or other metadata) encodes secret
- Graphics buffer attack:
 - Write to screen
 - Read back with OCR
 - Or adjust widget position / font size to encode secret

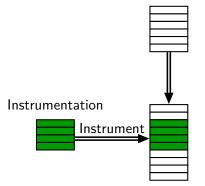
Summary

- Dynamic taint analysis tracks tainted values (from taint sources)
- Tags also referred to as shadow values
- Removes taint if values are sanitised
- Detects attempts to use tainted values in taint sinks
- Still many weaknesses in analysis:
 - Control-dependence attacks
 - System command attacks
 - Side-channel attacks
- Can be strengthened with symbolic techniques

Dynamic Binary Analysis

- Binary Analysis: Analyse binary executables
 - Applicable to any executable program
 - Only requires binary code
 - Unaware of source language
- Dynamic Binary Analysis
 - Analyser runs concurrently with program-under-analysis
 - Can adaptively instrument / analyse / intercede

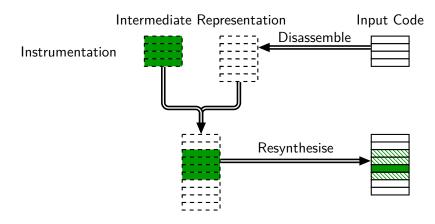
Dynamic Binary Instrumentation (1/3)



Input Code

Copy-and-Annotate

Dynamic Binary Instrumentation (2/3)



Disassemble-and-Resynthesise

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Dynamic Binary Instrumentation (3/3)

Copy-and-Annotate (e.g., pin):

- Inserts code into binary
- Inserted code must maintain state (registers!)
- Disassemble-and-Resynthesise (e.g., valgrind, qemu):
 - Decomposes program into IR
 - Instrumentation on IR-level
 - Easier/faster to track shadow values in some cases
 - Shadow registers
 - Shadow memory
 - Must model system calls for proper tracking

Application: Finding Memory Errors

- Reads from uninitialised memory in C can trigger undefined behaviour
- Approach: Track information: which bits are uninitialised?
- Requires shadow registers, shadow values
- Almost every instruction must be instrumented

. . .

Shadow values Program



short x; x |= 0x7; if (x & 0x10) {

Example: Valgrind's Memcheck

- Valgrind is Disassemble-and-Resynthesise-style Binary Instrumentation tool
- Memcheck: tracks memory initialisation (mostly) at bit level
 - Less precise for floating point registers
- Valgrind uses dynamic translation:
 - Translate & instrument blocks of code at address until return / branch
 - Instrumented code jumps back into Valgrind core for lookup / new translation

Challenges

- System calls
 - System calls may affect shadow values (e.g., propagate taintedness)
 - Must be modelled for precision
- Self-modifying code
 - ▶ Used e.g. in GNU libc
 - Must be detected, force eviction of old code (expensive checks!)

Valgrind

Valgrind

- Binary instrumenter
- Available platforms:
 - ▶ x86/Linux (partial) and Darwin
 - AMD64/Linux and Darwin
 - ▶ PPC64/Linux, PPC64LE/Linux (≤ Power8)
 - ► S390X/Linux
 - ARM(64)/Linux (≥ ARMv7)
 - MIPS32/Linux, MIPS64/Linux
 - Solaris
 - Android
- Analyses (focus on Simulation):
 - Call analysis
 - Cache analysis
 - Memcheck

Qemu



- Binary instrumenter and translator
- Focus on emulation
- Runs kernel + user space
- Translate from one ISA to another (e.g., run ARM on ADM64)
- Emulates system:
 - ▶ Graphics, networking, sound, input devices, USB, ...
- Almost two dozen platforms supported

Summary

- Binary instrumentation is a form of low-level dynamic analysis
- Two main schemes:
 - **Copy-and-Annotate**: insert new code
 - Disassemble-and-Resynthesise: merge analysis subject code with annotation code
- Shadow values supported through shadow registers and shadow memory