Cooperative Bug Isolation

Ben Liblit et al.
What’s This?

• I decided that that sigma calculus for objects was “too heavy” for our final lecture.
• OO slides are available on the webpage.
• Instead, we’ll talk about the work that won the 2005 ACM Doctoral Dissertation Award.
Sic Transit Gloria Raymondi

• Bugs experienced by users matter.

• We can use information from user runs of programs to find bugs.

• Random sampling keeps the overhead of doing this low.

• Large public deployments exist.
Today’s Goal: Measure Reality

• We measure bridges, airplanes, cars...
  - Where is flight data recorder for software?

• Users are a vast, untapped resource
  - 60 million XP licenses in first year; 2/second
  - 1.9M Kazaa downloads per week in 2004; 3/s
  - Users know what matters most
    • Nay, users *define* what matters most!

• Opportunity for *reality-directed* debugging
  - Implicit bug triage for an imperfect world
Bug Isolation Architecture

No annotation required; just pick what to instrument.

Program Source → Sampler → Compiler → Shipping Application

Guesses

Top bugs with likely causes

Statistical Debugging

Profile & 😊/😊
Why Will This Work?

• Good News: Users can help!
• Important bugs happen often, to many users
  - User communities are big and growing fast
  - **User runs vastly exceed testing runs**
  - Users are networked
• We *can* do better, with help from users!
  - cf. crash reporting (Microsoft, Netscape)
  - Today: research efforts
Let’s Use Randomness

- Problem: recording everything is too expensive!
- Idea: each user records 0.1% of everything
- Generic sparse sampling framework
  - Adaptation of Arnold & Ryder
- Suite of instrumentations / analyses
  - Sharing the cost of assertions
  - Isolating deterministic bugs
  - Isolating non-deterministic bugs

How do profilers work?
Sampling the Bernoulli Way

- Identify the points of interest
- Decide to examine or ignore each site...
  - Randomly
  - Independently
  - Dynamically

× Cannot use clock interrupt: no context
× Cannot be periodic: unfair
× Cannot toss coin at each site: too slow
Anticipating the Next Sample

- Randomized global countdown

- Selected from geometric distribution
  - Inter-arrival time for biased coin toss
  - Stores: How many tails before next head?
    - i.e., how many sampling points to skip before we write down the next piece of data?

- Mean of distribution = expected sample rate
Amortized Coin Tossing

- Each acyclic region:
  - Finite number of paths
  - Finite max number of instrumentation sites
  - Shaded nodes represent instrumentation sites
Amortized Coin Tossing

- Each acyclic region:
  - Finite number of paths
  - Finite max number of instrumentation sites
- Clone each region
  - “Fast” variant
  - “Slow” sampling variant
- Choose at run time
Optimizations

- Cache global countdown in local variable
  - Global $\rightarrow$ local at func entry & after each call
  - Local $\rightarrow$ global at func exit & before each call
- Identify and ignore “weightless” functions
- Avoid cloning
  - Instrumentation-free prefix or suffix
  - Weightless or singleton regions
- Static branch prediction at region heads
- Partition sites among several binaries
- Many additional possibilities ...
This 1991-1996 Nickelodeon cartoon featured the song "Happy Happy Joy Joy", a gilded yak, the catchphrase "what is it, man?" and a Slinky parody based on logs.
Q: Cartoons (668 / 842)

• Identify the speaker of and the next five words after "C is for Cookie".
Q: Music (195 / 842)

• Name any one of Simon and Garfunkel's rhyming Fifty Ways To Leave Your Lover.
This over-the-top 1992 Baz Luhrmann romantic comedy mockumentary features Australians Scott and Fran competing in the "Pan-Pacific Grand Prix".
Sharing the Cost of Assertions

• Now we know how to sample things.
• Does this work in practice?
  - Let’s do a series of experiments.
• First: microbenchmark for sampling costs!

• What to sample: `assert()` statements

• Identify (for debugging) assertions that
  - *Sometimes fail* on bad runs
  - But *always succeed* on good runs
Case Study: CCured Safety Checks

- Assertion-dense C code
- Worst-case scenario for us
  - Each assertion extremely fast
- No bugs here; purely performance study
  - Unconditional: 55% average overhead
  - $\frac{1}{100}$ sampling: 17% average overhead
  - $\frac{1}{1000}$ sampling: 10% average; half below 5%
Isolating a Deterministic Bug

• Guess predicates on scalar function returns
  \( f() < 0 \) \( f() == 0 \) \( f() > 0 \)

• Count how often each predicate holds
  - Client-side reduction into counter triples

• Identify differences in good versus bad runs
  - Predicates \textit{observed true} on some bad runs
  - Predicates \textit{never observed true} on any good run

Function return triples aren’t the only things we can sample.
Case Study: \texttt{ccrypt} Crashing Bug

- 570 call sites
- $3 \times 570 = 1710$ counters
- Simulate large user community
  - 2990 randomized runs; 88 crashes
- Sampling density $\frac{1}{1000}$
  - Less than 4% performance overhead
- Recall goal: sampled predicates should make it easier to debug the code ...
Winnowing Down to the Culprits

- 1710 counters
- 1569 are always zero
  - 141 remain
- 139 are nonzero on some successful run
- Not much left!

```
file_exists() > 0
xreadline() == 0
```

How do these pin down the bug? You’ll see in a second.
Isolating a Non-Deterministic Bug

- Guess: at each *direct scalar assignment* \( x = \ldots \)
- For each same-typed in-scope variable \( y \)
- Guess predicates on \( x \) and \( y \)
  \( (x < y) \quad (x == y) \quad (x > y) \)
- Count how often each predicate holds
  - Client-side reduction into counter triples
Case Study: \texttt{bc} Crashing Bug

- Hunt for intermittent crash in \texttt{bc-1.06}
  - Stack traces suggest heap corruption
- 2729 runs with 9MB random inputs
- 30,150 predicates on 8910 lines of code
- Sampling key to performance
  - 13\% overhead without sampling
  - 0.5\% overhead with $\frac{1}{1000}$ sampling
Statistical Debugging via Regularized Logistic Regression

- S-shaped cousin to linear regression
- Predict success/failure as function of counters
- Penalty factor forces most coefficients to zero
  - Large coefficient $\Rightarrow$ highly predictive of failure
void more_arrays ()
{
    ...

    /* Copy the old arrays. */
    for (indx = 1; indx < old_count; indx++)
        arrays[indx] = old_ary[indx];

    /* Initialize the new elements. */
    for (; indx < v_count; indx++)
        arrays[indx] = NULL;

    ...
}

#1: indx > scale
#2: indx > use_math
### Top-Ranked Predictors

void more_arrays ()
{
    …

    /* Copy the old arrays. */
    for (indx = 1; indx < old_count; indx++)
        arrays[indx] = old_ary[indx];

    /* Initialize the new elements. */
    for (; indx < v_count; indx++)
        arrays[indx] = NULL;

    …
}

#1: indx > scale
#2: indx > use_math
#3: indx > opterr
#4: indx > next_func
#5: indx > i_base
Bug Found: Buffer Overrun

```c
void more_arrays ()
{
    ...

    /* Copy the old arrays. */
    for (indx = 1; indx < old_count; indx++)
        arrays[indx] = old_ary[indx];

    /* Initialize the new elements. */
    for (; indx < v_count; indx++)
        arrays[indx] = NULL;

    ...
}
```
Moving To The Real World

- Pick instrumentation scheme
- Automatic tool instruments program
- Sampling yields low overhead
- Many users run program
- Many reports $\Rightarrow$ find bug
- So let’s do it!
Multithreaded Programs

• Global next-sample countdown
  - High contention, small footprint
  - Want to use registers for performance
  ⇒ Thread-local: one countdown per thread

• Global predicate counters
  - Low contention, large footprint
  ⇒ Optimistic atomic increment
Multi-Module Programs

• Forget about global static analysis
  - Plug-ins, shared libraries
  - Instrumented & uninstrumented code

• Self-management at compile time
  - Locally derive identifying object signature
  - Embed static site information within object file

• Self-management at run time
  - Report feedback state on normal object unload
  - Signal handlers walk global object registry
Native Compiler Integration

- Instrumentor must mimic native compiler
  - You don’t have time to port & annotate by hand
- This approach: source-to-source, then native
- Hooks for GCC:
  - Stage wrapping via scripts
  - Flag management via specfiles
Keeping the User In Control

The Cooperative Bug Isolation Project

Some applications on this computer can monitor their own behavior while they run. Each time you use a participating application, you can help to make it better for everyone.

Feedback from users like you can help us find and fix the bugs that matter most. Do you wish to provide automatic feedback when you use participating applications on this computer?

- Yes, count me in
- No thank you

If you choose “Yes,” then participating applications will send feedback to the bug isolation center after each run. Failed runs will also include crash reports to help us see what went wrong.

If you choose “No,” then participating applications will not monitor their own behavior. No automatic feedback will ever be sent, though you can still report problems manually.

Not sure what to do? Click here to learn more.
Public Deployment 2004

- Evolution
- Gaim
- GIMP
- Gnumeric
- Nautilus
- Rhythmbox

Reports Received

Good
Error
Crash
Public Deployment 2004

- Evolution
- Gaim
- GIMP
- Gnumeric
- Nautilus
- Rhythmbox

Graph showing deployment percentages with categories for Good, Error, and Crash.
Sneak Peak: Data Exploration
Summary: Putting it All Together

• Flexible, fair, low overhead sampling

• Predicates probe program behavior
  - Client-side reduction to counters
  - Most guesses are uninteresting or meaningless

• Seek behaviors that co-vary with outcome
  - Deterministic failures: process of elimination
  - Non-deterministic failures: statistical modeling
Conclusions

• **Bug triage that directly reflects reality**
  - Learn the most, most quickly, about the bugs that happen most often

• **Variability is a benefit rather than a problem**
  - Results grow stronger over time

• **Find bugs while you sleep!**

• **Public deployment is challenging**
  - Real world code pushes tools to their limits
  - Large user communities take time to build

• **But the results are worth it:**

  “Thanks to Ben Liblit and the Cooperative Bug Isolation Project, this version of Rhythmbox should be the most stable yet.”
Homework

- Project writeups due today …
- Good luck with your project presentations!