Lecture 14: Testing Strategies for Code Debugging 36-350 13 October 2014

Last Time: Basic Debugging

Basic tricks for debugging:

- Notifications and alerts that you can add
- Localizing issues and changing input parameters
- Precomputed results

Today: Intermediate Debugging

Better success through design!

- Trusting our results through modular design
- Building tests: functional tests (top-level), unit tests (bottom-level)

Procedure versus Substance

Our two competing goals:

- Do we get the right answer (substance)?
- Do we get an answer in the right way (procedure)?

An important distinction, as these these go back and forth with each other:

- We trust a procedure because it gives the right answer.
- We trust the answer because it came from a good procedure.

Since programming means *making* a procedure, we check the substance primarily.

Testing for particular cases

Test cases with known answers

```
add <- function (part1, part2) { part1 + part2 }
a <- runif(1)
add(2,3) == 5</pre>
```

[1] TRUE

add(a,0) == a

[1] TRUE

add(a,-a) == 0

[1] TRUE

Testing for particular cases

Real numbers and floating-point precision

cor(c(1,-1,1,1),c(-1,1,-1,1))

[1] -0.5774

-1/sqrt(3)

[1] -0.5774

cor(c(1,-1,1,1),c(-1,1,-1,1)) == -1/sqrt(3)

[1] FALSE

Testing by cross-checking

Compare alternate routes to the same answer:

```
test.unif <- runif(n=3,min=-10,max=10)
add(test.unif[1],test.unif[2]) ==
  add(test.unif[2],test.unif[1])</pre>
```

[1] TRUE

```
add(add(test.unif[1],test.unif[2]),test.unif[3]) ==
add(test.unif[1],add(test.unif[2],test.unif[3]))
```

[1] TRUE

```
add(test.unif[3]*test.unif[1],test.unif[3]*test.unif[2]) ==
test.unif[3]*add(test.unif[1],test.unif[2])
```

[1] FALSE

wwwwww ## Testing by cross-checking

Test function: numerical derivative

```
x <- runif(10,-10,10)
f <- function(x) {x^2*exp(-x^2)}
g <- function(x) {2*x*exp(-x^2) -2* x^3*exp(-x^2)}
isTRUE(all.equal(derivative(f,x), g(x)))</pre>
```

Testing by cross-checking

If this seems too unstatistical...

xx <- runif(10) aa <- runif(1) cor(xx,xx) == 1 ## [1] TRUE cor(xx,-xx) == -1 ## [1] FALSE cor(xx,aa*xx) == 1

[1] TRUE

Testing by cross-checking

```
pp <- runif(10); mean=0; sd=xx
all(pnorm(0,mean=mean,sd=sd) == 0.5)</pre>
```

[1] TRUE

```
pnorm(xx,mean,sd) == pnorm((xx-mean)/sd,0,1)
```

Testing by cross-checking

```
all(pnorm(xx,0,1) == 1-pnorm(-xx,0,1))
```

[1] TRUE

pnorm(qnorm(pp)) == pp

[1] FALSE TRUE TRUE TRUE FALSE TRUE TRUE TRUE TRUE TRUE

qnorm(pnorm(xx)) == xx

[1] TRUE FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE

With finite precision we don't really want to insist that these be exact!

Software Testings vs. Hypothesis Testing

Statistical hypothesis testing: risk of false alarm (size) vs. probability of detection (power) – this balances type I vs. type II errors

In software testing: no false alarms allowed (false alarm rate = 0). This must reduce our power to detect errors; code can pass all our tests and still be wrong.

But! we can direct the power to detect certain errors, *including* where the error lies, if we test small pieces.

Combining Testing and Coding

The idea behind unit testing:

- A variety of tests gives us more power to detect errors, more confidence when tests are passed.
- By breaking code into self-enclosed functions, we can better identify problems.
- *Therefore*: for each function, we build a battery of tests that are easy to step through and identify problems.
- This makes it easier to add new tests to a function as well.
- By bundling these tests into their own function, we keep program flow clean and remind ourselves later why this mattered!

The Great Testing Cycle

After making changes to a function, re-run its tests, and those of functions that depend on it.

- If anything's (still) broken, fix it; if not, continue.
- When you meet a new error, write a new test.
- When you add a new capacity, write a new test.

A Ratchet Approach: "Regression Testing"

When we have a version of the code which we are confident gets some cases right, keep it around (under a separate name).

- Now compare new versions to the old, on those cases
- Keep debugging until the new version is at least as good as the old

Test-Driven Development

General strategy for development.

- 1. Have an idea about what the program should do.
 - Idea is vague and unhelpful
 - Make it clear and useful by writing tests for success
 - Tests come *first*, then the program
- 2. Modify code until it passes all the tests

- 3. When you find a new error, write a new test
- 4. When you add a new capacity, write a new test
- 5. When you change your mind about the goal, change the tests
- 6. By the end, the tests specify what the program should do, and the program does it

Awkward Cases

Boundary cases, "at the edge" of something, or non-standard inputs. Such as:

[1] NA

add(5,NA)

```
try(add("a","b")) # NA, or error message?
divide <- function (top, bottom) top/bottom
divide(10,0) # Inf, presumably
```

NA, presumably

[1] Inf

divide(0,0) # NA?

[1] NaN

Awkward Cases

Pinning down awkward cases helps specify function

var(1) # NA? error?

[1] NA

cor(c(1,-1,1,-1),c(-1,1,NA,1)) # NA? -1? -1 with a warning?

[1] NA

```
try(cor(c(1,-1,1,-1),c(-1,1,"z",1))) # NA? -1? -1 with a warning?
try(cor(c(1,-1),c(-1,1,-1,1))) # NA? 0? -1?
```

Pitfalls

- Writing tests takes time
- Running tests takes time
- Tests have to be debugged themselves
- Tests can provide a false sense of security
- There are costs to knowing about problems (people get upset, responsibility to fix things, etc.)

Summary

- Trusting software means testing it for correctness, both of substance and of procedure
- Software testing is an extreme form of hypothesis testing: no false positives allowed, so any power to detect errors has to be very focused
- \therefore Write and use lots of tests; add to them as we find new errors
- Cycle between writing code and testing it