Analysis of Test Coverage Data on a Large-Scale Industrial System

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Background Required Theory

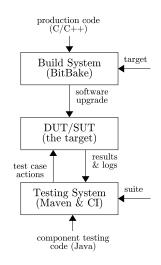
- Importance of software testing, within any mission critical software products.
- See for e.g. NASA's Climate Orbiter...
- The code/test coverage metric can be used in determining e.g. testing holes.

```
1 int factorial(int n) {
2     if (n == 0) return 1;
3     else if (n < 0) return -1;
4     return factorial(n - 1)*n;
5 }
6
7 int main(int, char**) {
8     assert(factorial(0) == 1);
9     assert(factorial(4) == 24);
10     return 0;
11 }</pre>
```



¹ Climate Orbiter: http://mars.nasa.gov/msp98/

- Project at Ericsson (R&D) Linköping, the given large-scale industrial system.
- Production & unit tests: 960 KSLOC, the target function tests: 564 KSLOC.
- Tasks: integrate a coverage gathering, and analysis system for function tests; determine "feasibility" of the addition.
- Goal: develop framework, upon which test case selection/prioritization could be derived in future work. PL3 ≈ 18h.



Motivation

- Software testing is complicated and expensive, research [B⁺81] has shown 50% project effort.
- Coverage data provides insight into the nature of tests; giving useful information about them.
- Research in *large-scale test coverage is scarce*, according to several papers such as [ABR⁺11]
- Gathering & analysis is problematic [ABR+11]
- Articles don't present *development experience*.

Problems

- Coverage gathering is problematic on large-scale systems:
 - Existing coverage tools *don't integrate well* on all setups.
 - Performance and resource usage might be neg. affected.
 - Research question: how feasible is such an extension?
- Analyzing raw test coverage data manually isn't feasible:
 - Huge amounts of data produced for large-scale software.
 - Difficult to extract any meaningful test case properties.
 - Research question: what does the analysis tools find?

Challenges

- Performance sensitive system.
- Executes a series of *daemons*.
- Coverage for a *remote target*.
- Huge amount of profile data.
- Isolate changes in behaviour.



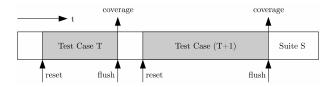
```
1 static int bb[4] = {};
  static int factorial(int n) {
 3
       ++bb[0];
       if (n == 0)
           ++bb[1];
           return 1;
 7
       else if (n < 0)
           ++bb[2];
 8
 9
           return -1;
10
11
12
       ++bb[3];
13
       return factorial(n - 1)*n;
14 }
15
16 int main(int, char**)
17
       assert(factorial(0) == 1);
18
       assert(factorial(4) == 24);
19
       printf("bb 0: %i", bb[0]);
20
       printf("bb 1: %i", bb[1]);
21
       printf("bb 2: %i", bb[2]);
22
       printf("bb 3: %i", bb[3]);
23
       return 0;
24 }
```

- Instrument production code with coverage capabilities.
- Measure the performance effects of these, statistically.
- Add *flushing signal*, to deal with the *daemon software*.
- Extend testing system to fetch coverage on test's end.
- Build/find tool to analyze, and locate test similarities.

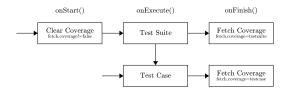
- Production code is built with GCC → GCov is built-in, & proven to have minimal performance impact, ≈ 3%.
- e.g.: -fprofile-arcs -ftest-coverage -00.
- However, remote target → coverage is dumped wrong, solved with GCOV_PREFIX & GCOV_PREFIX_STRIP.
- Enabled in the *build system* with: --coverage flag. (←)

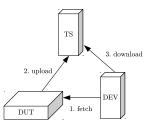
Solution Daemon Flushing

- Several instrumented daemons on the target device.
- Coverage information is dumped when terminated, however, device *reboots* if this is done... Not good.
- By calling __gcov_flush(), processes can dump, without being terminated explicitly (no rebooting).
- Synchronize flush: kill -sUSR1 \$(pidof *), handler enabled on all daemons by: --coverage. (←)



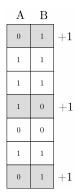
- Extends the Maven/Jcat testing flow.
- Automatically flushes coverage down, splitting the individual test case data.
- Switched with fetch.coverage = (testcase | testsuite). (←)





- How to measure test case similarities? see Jaccard index, Hamming distance.
- Initial solution modified gcov-tool.
- Current implementation scovat.py, gives set operations on coverage data, required by Hamming & Jaccard. (←)

$$J(A,B) = rac{|A \cap B|}{|A \cup B|}, \ d_J = 1 - J(A,B)$$

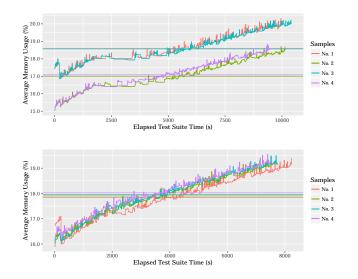


- Build software with bb --coverage <recipe>, enabling coverage instrumentations and daemon signal handlers setup.
- 2 Upload software packages with stp_set_up <deviceid>.
- 3 Enable the fetch.coverage=testcase testing property, enabling the Maven/JCat testing framework fetch coverages.
- 4 Execute the desired test case/suite, e.g. PL2 on the devices.
- 5 Test coverage data is continuously fetched to the developer.
- **6** scovat.py -gb \$BUILDLOC -o cache generates the intermediate coverage format, used later for modifying data.
- 7 scovat.py -ao analysis cache/<A> cache/ gives Jaccard coefficient, and Hamming distance for (A, B).

- Software *instrumentation* haven't made any tests unstable.
- Fetching coverage from target → developer adds 20s time, largely caused by bugs in the *Trilead SSH* implementation.
- Consumes a *fixed disk space* of 8.3 MiB for each test case, which is transferred & removed continuously when fetched.
- The suite execution time, processor, and memory usage are measured statistically after a t-distribution, α = 5%, n = 4.

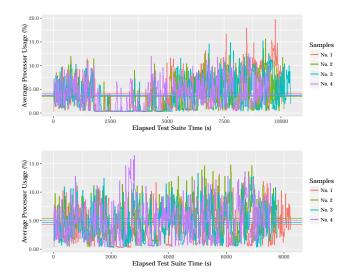
Sampled Dataset	Lower Time (h)	Upper Time (h)	
Instrumented	2.656	2.911	
Non-Instrumented	2.075	2.260	

Results Memory Usage



Sampled Dataset	Lower Usage (%)	Upper Usage (%)	
Instrumented	16.7663	18.8446	
Non-Instrumented	17.8428	18.0185	

Results Processor Usage



Sampled Dataset	Lower Usage (%)	Upper Usage (%)	
Instrumented	3.43483	4.12369	
Non-Instrumented	4.28527	5.40044	

- Retrieved from the primary development test suite.
- Upon producing *lcov report* with fetched coverage:
 - Statement coverage: 59.3% out of 96158 (lines).
 - Function coverage: 70.7%, of 23 870 (functions).
 - Branch coverage: 24.6%, (retrieved from: Gcov).

- Demonstration of *test similarity analysis*, with 3 tests:
 - A: IPForwarding#testCliRejectsInvalidAddressOnDstMo,
 - B: IPForwarding#testCliRejectsInvalidAddressOnNexthopMo,
 - **C:** PL1#testPL1TestSuite, *all three from Ericsson's tests*.
- Both **A** and **B** should exercise very similar code locations.
- While C exercises more varied locations, different from A.
- Similarity leads to potential for test redundancy [CNM07].
- Note!: exercising similar locations ⇒ exactly same tests!

Criterion	$d_H(A,B)$	A ∩ B	A ∪ B	J(A, B)
Statement	0	400	400	1.00000
Function	0	90	90	1.00000
Branch	0	132	132	1.00000
Criterion	$d_H(A,C)$	A ∩ C	A ∪ C	J(A, C)
Statement	21 691	398	22 089	0.01801
Function	5 369	90	5 459	0.01648
Branch	21 960	123	22 083	0.00557

Conclusions

- Feasibility: deemed possible, since tests aren't unstable; however, it increases softw. execution time significantly.
- Measurements: project has similar coverage to Google's average project (C) coverage (statement) measurement.
- Interpretation: analysis tool/method, show locations of: test case similarity, and pot. test redundancy, [CNM07].
- Limitation: still requires engineers to verify redundancy.
- Future Work: clustering, test selection & prioritization.

- Software testing is hard; test coverage give us valuable metadata about tests.
- Gathering & analyzing code coverages on large-scale systems proves difficult.
- Integration of the system was feasible, but prog. execution time was affected.
- Metrics of a real, large-scale test suite were given since such data was scarce.
- Finally, analysis tool shows similarities between test cases, which sifts results.



THE BEST THESIS DEFENSE IS A GOOD THESIS OFFENSE

³ Thesis Defense: https://www.xkcd.com/1403/

Questions?

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