Test Automation

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The problem of test automation

Testing has repetitive components, so automation is justified The problem is *cost-benefit evaluation* of automation [Kaner]

Time for: test creation, checking their functionality, documentation Is automation reusable? (if the program evolves)

Is maintenance needed? (GUI change, internationalization)

⇒ Test automation: treated like any software development

Does it delay finding bugs? (fewer resources to run tests)

Does it find enough bugs? Or are most found by manual testing

Is it powerful enough? Or does it automate only "easy" tests?

Example: Capture-replay

- 1) Record user actions (mouse/keyboard) and resulting screen (bitmap) \Rightarrow most primitive level
 - other checks: with tester effort (interrupt/insert)
 - fragile: susceptible to any product change
 - possible comparison errors in resulting image
- 2) script with high-level actions (select menu/button)
- more flexible, but does not check graphic layout (low level: font, text size/overwrite, etc.)
- 3) scripting language to automatically generate new tests

Disadvantage of capture-replay

Cannot continue from errors

- ⇒ errors are found manually in the recording process
- ⇒ only rerunning a "good" test is automated (regression)

Does not define tests *implicit* for human ("all the rest is OK") (cannot detect unspecified errors, is inflexible – e.g. bitmap)

Example: Test monkeys

Automated tools that execute random tests (without a testor's knowledge on product functionality) Dumb monkeys: completely ignore purpose (know just mouse/keyboard)

but may have basic notions about windows/menus/buttons *Smart monkeys*: have a *state model* of the application, explore transitions between these states

- ++ can sometimes find 10-20% of errors [Nyman, Microsoft, 2000]
- ++ good preliminary coverage (e.g.: 65% in 15 min for a text editor)
- ++ completely automated, no human effort for test capture
- -- "dumb": only bug known to monkeys is system crash
- $-- \Rightarrow$ errors are hard to record and reproduce
- ++ runs independently, unsupervised, minimal resources (cost)

What can we automate in testing?

Test execution

e.g. any unit testing framework useful in

Test evaluation

- = problem of *test oracle* : did the test pass ? Nontrivial, often needs manual inspection. Risks:
- undetected errors (imprecision)
- false warnings ⇒ cost of manual checking

Ex: compare continuous signals (in automotive industry) image comparison (for screen/printer)

Test generation

Relatively easy: generating test skeletons (declarations + calls) More difficult: intelligent generation of relevant data (coverage)

Choosing a test architecture [Kaner]

- 1) Data-driven architecture
 separates data from test structure (like in programs)
 Example: table. row = test; columns = test parameters
 A script generates a test case for every table row
 Minimal reasonable coverage: every pair of parameter values
 (for every combination of values, number is exponential)
- 2) Framework-based architecture

A library of functions separates testing from UI

- e.g. open(file), independent of actions for opening (menu, button click, keyboard, etc.)
- ++ reuse for frequent actions
- ++ indirection \Rightarrow insulation from testing tool
- -- costly, amortized only in future releases

Specification-based testing

Automatable (*keyword testing*) for spects in well-defined language Starting from documentation: tabular spec, e.g. [Pettichord]

Test ID	Operation	Table	Name	Туре	Nulls
dtbed101	Add Col	TB03	NEW_INT_COL	CHAR(100)	Υ

Important: choose format easily understandable by user A translator/test interpreter generates the test driver from the table or interfaces with the (commercial) testing tool used ++: requirement-driven (*what*, not *how*), independent of implementation and testing tool, self-documented

More advanced: automated test generation from specs in formal language

e.g. decision tables in RSML in TCAS-II aviation protocol test generation from timing diagrams in embedded systems

Model-based testing

Models: finite automata, UML, Statecharts (hierarchical automata), Message Sequence Charts, timed automata, Petri nets, Markov chains...

Test generation criteria: satisfactory model coverage all states / transitions; combinations of k consecutite transitions $(k\text{-}switch\ cover})$

- ++ facilitates generation of relevant tests
- -- investment in model building and maintenance

Testing based on *model checking* by *state space exploration*, starting from specifications:

- 1) question: can the model reach a given state?
- 2) if so, a *model checker* will generate an example trace = test case

Implementation-based testing: symbolic execution

Goal: exercising program, satisfying a *coverage criterion* ⇒ needs: instrumentation to measure test coverage How: set of *paths*: random choice + directed search (to reach branches not yet covered)

Symbolic execution: executing program using expression with symbolic variables, rather than concrete (numeric) values Symbolic execution gathers path conditions for followed branches

Satisfiability of conditions is checked with specialized tools (satisfiability checkers, constraint solvers)

 \Rightarrow generate input data that will exercise that path or prove path is infeasible \Rightarrow stops exploring that path

Symbolic execution for program testing

described as early as 1976 (James C. King)

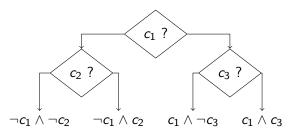
program is executed by a special interpreter, using *symbolic* inputs ⇒ results in symbolic execution tree traversal stops when path condition becomes unsatisfiable

Test generation purpose: attaining high coverage sometimes, reaching a specific branch

Successful mature technique, hundreds of papers, many tools: Java Pathfinder, (j)CUTE, C_{REST} , KLEE, Pex, SAGE, ... for C/C++, C#, Java, more recently JavaScript

Variants of symbolic execution

Classic *completely symbolic execution* explores each execution path independently



Problem: must express all program/language semantics as formula solving arbitrary formulas impossible (limited to simple arithmetic) reality: complex math, library function, environment

solution: model libraries & environment

e.g. KLEE tool has models for some 40 syscalls (2.5 kloc)

Dynamic (concolic) symbolic execution

symbolic execution is directed by concrete run (hence: "concolic"

When symbolic execution is infeasible, perform a concrete execution step

e.g. nonlinear arithmetic, library/system functions

```
function explore(pathcond = [c_1, c_2, ..., c_n])
for k = n downto 1 do

inputs = solve pathcond = c_1 \land ... \land c_{k-1} \land \neg c_k (flip c_k)
rerun with new inputs; capture new pathcond'
explore(pathcond')
```

Problem: by using concrete values, might not reach desired path

Concretization as potential obstacle

```
y = hash(x); // can't solve hash formula \Rightarrow y is
if (x + y > 0)
 // path 1
else
  // path 2
Assume: x = 20; y = hash(20) = 13 \Rightarrow path 1
To reach path 2, negate x + y > 0, with concrete y (constant 13)
Solver might return, e.g., x = -15
but we might have hash(15) = 27 (can't predict) and then
x + v > 0
  \Rightarrow execution still follows path 1
⇒ retry; worst-case: degrades to random testing
```

When is test automation efficient?

In *regression testing*: need only store tests and expected results (and means to automate comparison)

Testing user interfaces (discussed earlier)

Testing compilers / translators

automated test generation starting from input grammar explores random/statistic combination of grammar rules/nonterminals

Load/stress testing: random; quantity rather than content is relevant Fuzz testing: generate large quantities of random / possibly hostile input, to detect input validation errors or security vulnerabilities

e.g. $\rm RANDOOP$ [Microsoft]: 4 M tests in 150 CPU hours / 15 person-hr 30 bugs in code tested for 200 person-years, vs. 20 errors/year found manually

see also http://research.microsoft.com/en-us/projects/Pex/

Basic workings of a fuzzer

e.g. American Fuzzy Lop http://lcamtuf.coredump.cx/afl/maintains queue of test inputs mutates inputs using several strategies if new coverage achieved, add mutant to input queue minimize each test input (keeping coverage) minimize input corpus (avoids overlap)

records *transition coverage* between program basic blocks classifies runs into crashes/hangs/normal exit

highly successful, found many security vulnerabilities mutating inputs can eventually synthesize interesting formats (e.g. images)

can identify format fields with various meaning (length, checksum, payload, control opcode, etc.)

Automated debugging

After automating detection \Rightarrow help in fault localization

Minimizing test inputs binary search, finds input half that caused error (e.g. for file inputs)

Minimizarea diferențelor între o rulare corectă și una eronată tot căutare binară, pentru două intrări cât mai apropiate

Fault localization in space

in debugger, compare execution state between correct and buggy run

detect (precisely or statistically) invariants/patterns violated by erroneous run

Fault localization in time

compare erroneous runs and find points where ${\it infected}$ variables start affecting output

Delta debugging [Zeller]: partial automation of these techniques