Automated Exploratory Testing with Defined Coverage

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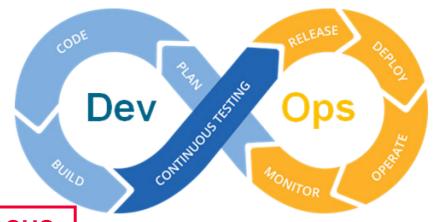
Challenges in continuous testing

- Lack of testability support in products
 - Required to enable automated testing
- Lack of standard tools
 - In-house test automation still common practice
- Lack of faster feedback loops

FOCUS

- Need to gather feedback from tests in real-time; short test execution time
- Lack of testing infrastructure
 - Test environments need to run 24/7 and under repeatable conditions
- Test data management
 - Centrally managed test data to get consistent test results
- Scalability issues
 - Due to complex systems and the need for large concurrent test sessions

Source: Richa Agarwal, Common Challenges in Continuous Testing, August 24, 2020



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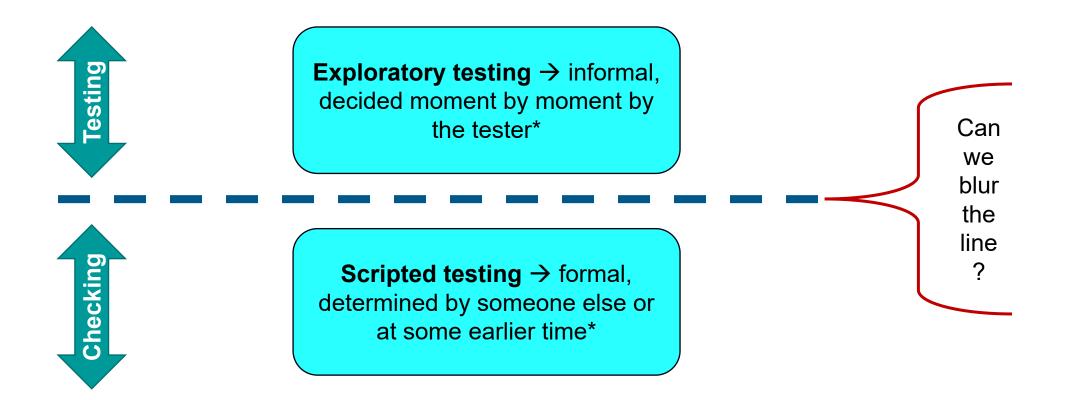
Exploratory testing

James Bach: "Testing is an exploratory process. It's not just sometimes exploratory; it is inherently exploratory."

Exploratory Testing is defined as simultaneous

- Learning,
- Test design and
- Test execution.

Testing vs Checking



* James Bach, https://www.satisfice.com/exploratory-testing





Adaptive testing to automate exploratory tests

- Testing of a self-contained software component
- Only the component API is accessed → black-box test
- Scenario tests
 - Structured sequence of predefined test commands
- Tester reacts to SUT responses at runtime
 - Selection of next test command according to an overall test goal + randomization

```
Given: Set of test commands
var tcmd = Reset();
while(true)
{
   tcmd.Invoke();
   tcmd = SelectNext();
}
```



Test commands

A test command = SUT interaction + local state update; it follows the "4-As" pattern (extension of "3-As" as known in unit testing)

- Arrange: prepare input parameters for SUT call
- Act: perform SUT call
- Assert: validate correctness of returned data
- Adapt: update local state in tester

A test command is conditioned

- A condition describes the state that enables the test command
- Finding the right condition and state representation is a creative, exploratory act
- Test commands form a guarded command language

Example: Stack [Condition(true)] TPush() { sut.Push(x); assert(sut.Length, i+1); i++; } [Condition(i > 0)] TPeek() { sut.Peek(); assert(sut.Length, i) } [Condition(i > 0)] TPop() { x = sut.Pop();assert(sut.Length, i-1); i--; }



Adaptive testing applied in practice

Test goal: coverage over test commands

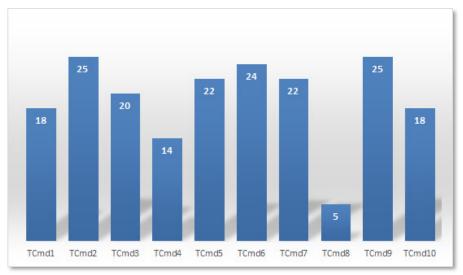
When is the set of test commands covered?

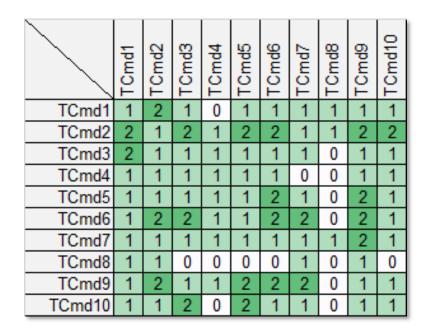
- Syntactical coverage over test command definitions
 → Counting test commands, pairs of test commands etc.
- Semantical coverage over states reached in test exec.

 \rightarrow Counting states

In practise, different goals are chosen to configure about 5 test runs lasting 10 min or less

- Each test run is different due to randomization
- Able to detect deep state failures
- Remain effective over the entire DevOps cycle





What difference does it make – Example 1: Deep state failures

- 5 adaptive tests detected a failure that 3453 unit tests and 1084 hard-coded integration tests were unable to find!
- Adaptive tests remain effective in finding failures over the entire development cycle

Test

Adaptive Tests

Integration Tests

Unit Tests

unit tests and integration e to find! main effective over the entire le		1	2			No So	~55 P	crass Toker	25 Propr	10 × 0 × 0 ×
Date/T	ïme	Duration	Duration Average	Passed	Failed Fixtures	Failed	Run		-	
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2017-0 03:07:0	09-20 09	1h 7m 44.525s	1s 177ms	3453	0	0	3453	*		
2017-0 01:59:		1h 42m 54.007s	5s 696ms	1084	0	0	1084	-t-		
	Summary	2h 58m 42.055s	2s 360ms	4538	0	5	4543	62		

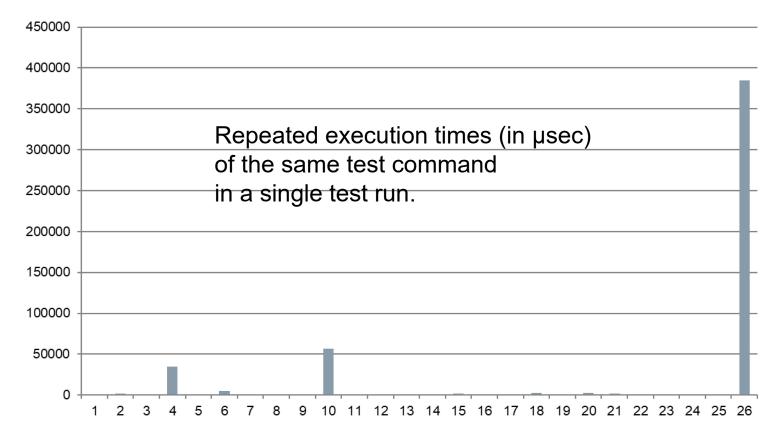
What difference does it make – Example 2: Performance degradation

Tracking of the execution time of test commands

- Within the same test run
 → Reliability tests
- Over multiple test runs over time
 - \rightarrow Regression tests

Expectation on execution times

- Constant
- Proportional
- Learned

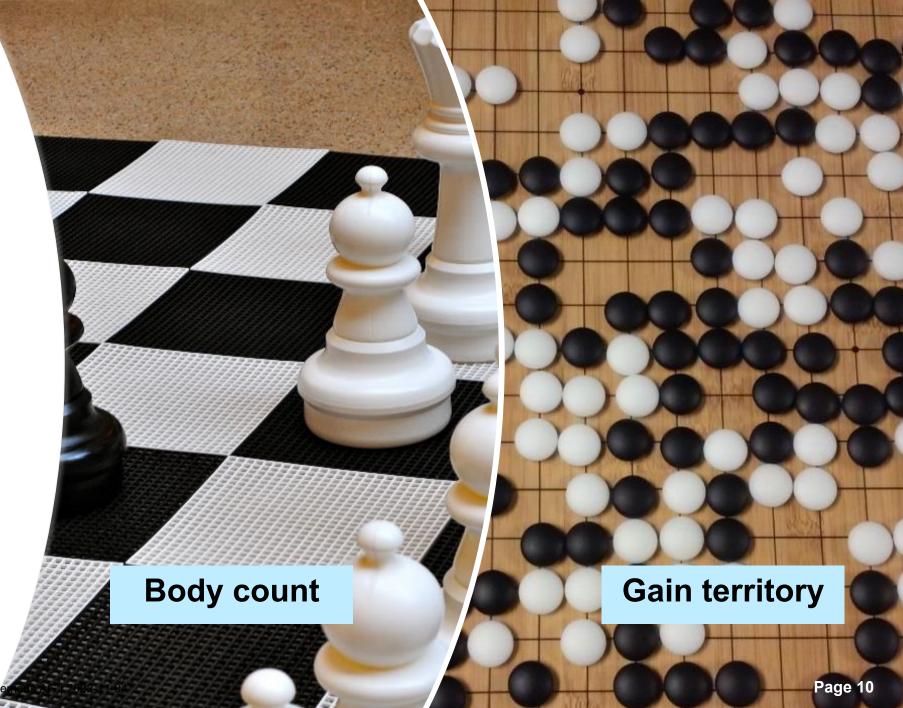


Common strategies in testing

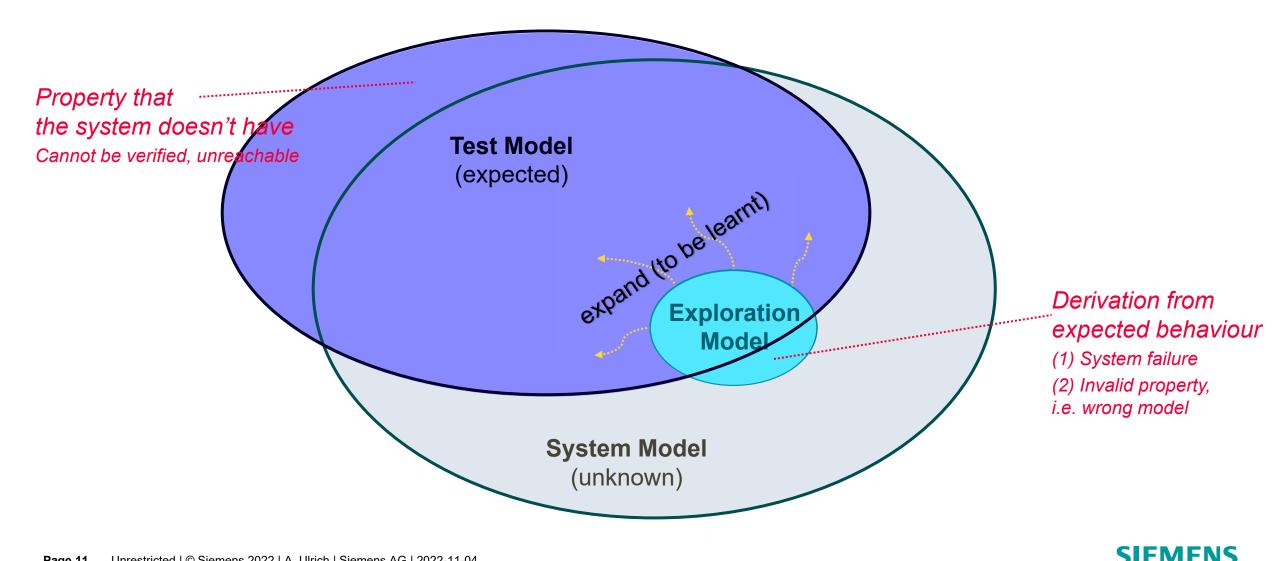
Detect failures until resources spent

Vs

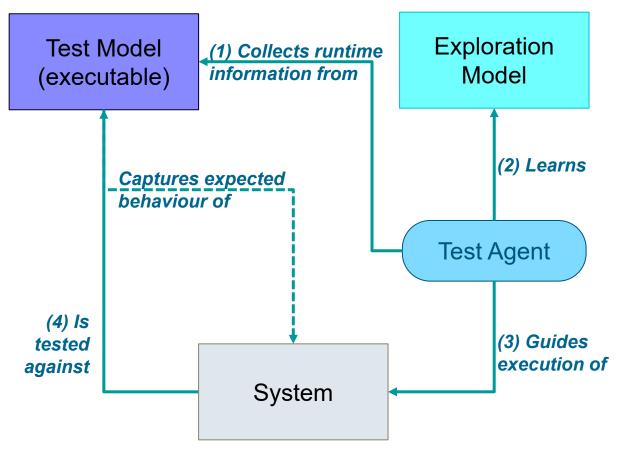
Gain confidence through covered behaviour



Test Model, System Model, and Exploration Model



Adaptive testing with learning



Model concepts

- Test model → set of test commands capturing system properties
- Exploration model \rightarrow is learnt from executions

Test goals

- Provide evidence that properties are satisfied
- Maximise coverage of exploration model

Test execution

- 1. Collect information about test commands
- 2. Learn model from executed test commands
- 3. Decide at runtime about next test commands
- 4. Run test commands against system

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Learning and inference of state machines from test execution traces

Trace: Execution sequence of inputs and outputs

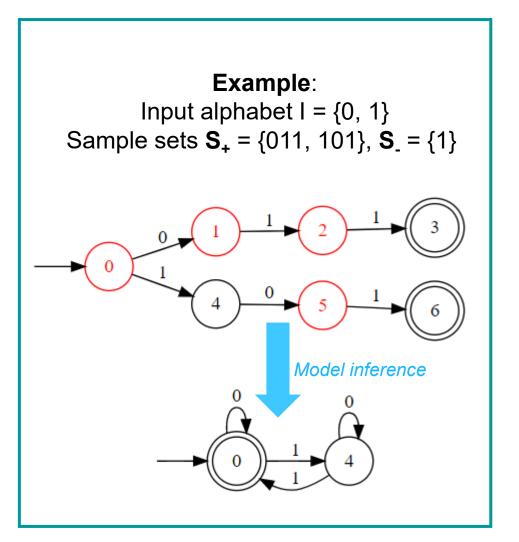
- I/O trace: sequence of input/output pairs
- Positive/negative samples: input sequence with evaluation of reached output
- Set of traces form a Prefix Tree

Inferred machine reproduces all traces from the training set with fewer states

- I/O traces \rightarrow Moore machine
- Pos/neg. samples \rightarrow DFA

Learning during the DevOps cycle, continuously

- Passive → Collect example traces randomly
- Active → Collect traces and derive queries





Passive learning and machine inference

Approach

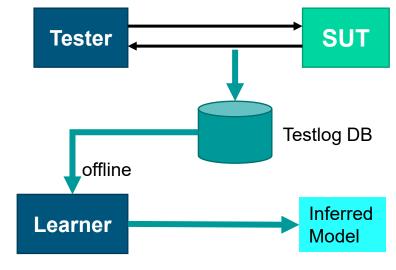
- Database to store test logs from test execution runs
- Test logs interpreted as I/O traces to construct a prefix tree
- Moore machine is inferred from the prefix tree

Adequacy criterion for machine inference

- Preferred: "Identification in the limit" All traces <u>longer than n</u> can be produced from equivalent machines
- Realistic:

All traces <u>up to length *n*</u> can be produced from equivalent machines

• Approximations to the minimum solution might be good enough

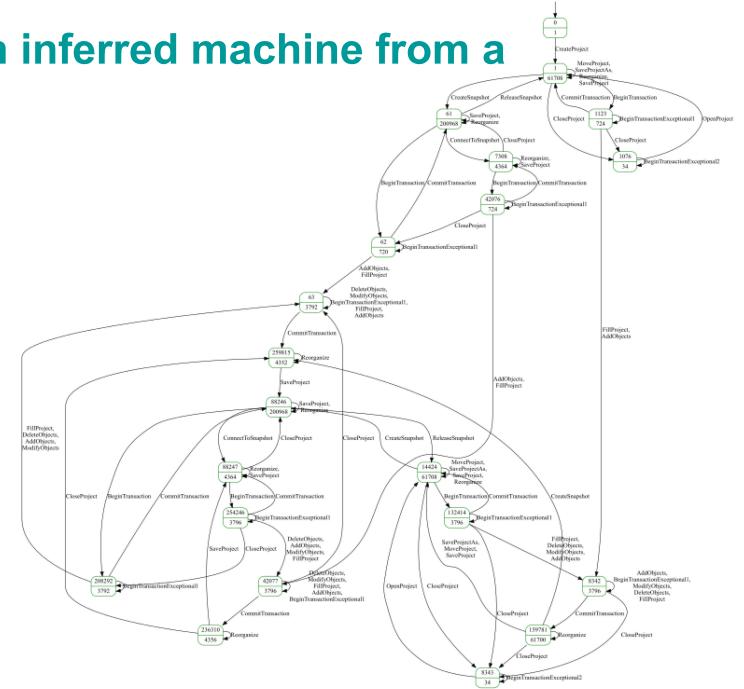


Passive Learning



Practical example of an inferred machine from a SW component test reateSnapshot 61 SaveProject,

- Inference from 544 test runs
- Building prefix tree in 0.754 sec
- Size of prefix tree is 275,881 states
- Inferring Moore machine in 11.312 sec
- Using 682.2 MB of heap memory
- Size of inferred machine is 21 • states



Conclusions

Exploratory testing doesn't need to stop at scripted tests

Adaptive testing preserves the exploration capability during runtime

Combining adaptive testing with learning to reach a defined level of state coverage

Learning helps find optimal solutions to a given coverage goal utilising:

- Exploration of new SUT behaviour
- Exploitation of learned knowledge