

CMPT 473
Software Testing, Reliability and Security

Property Based Testing

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Recall Unit Tests

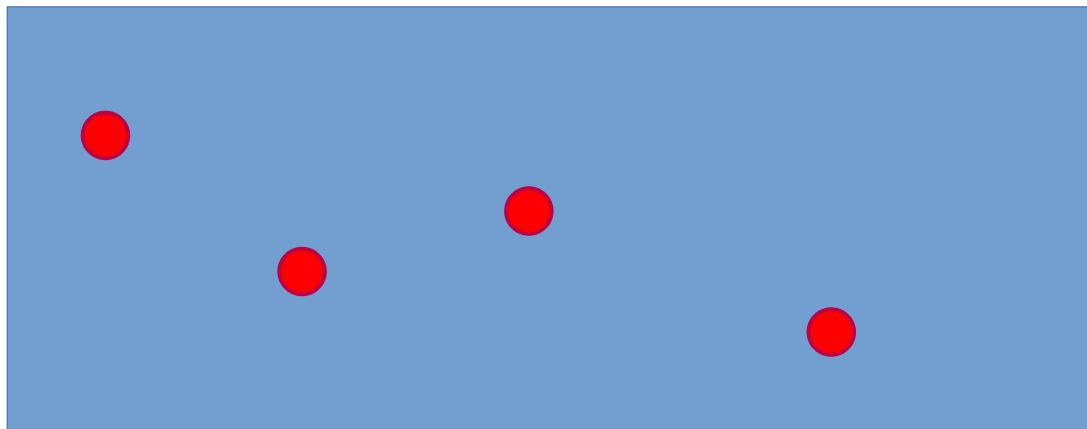
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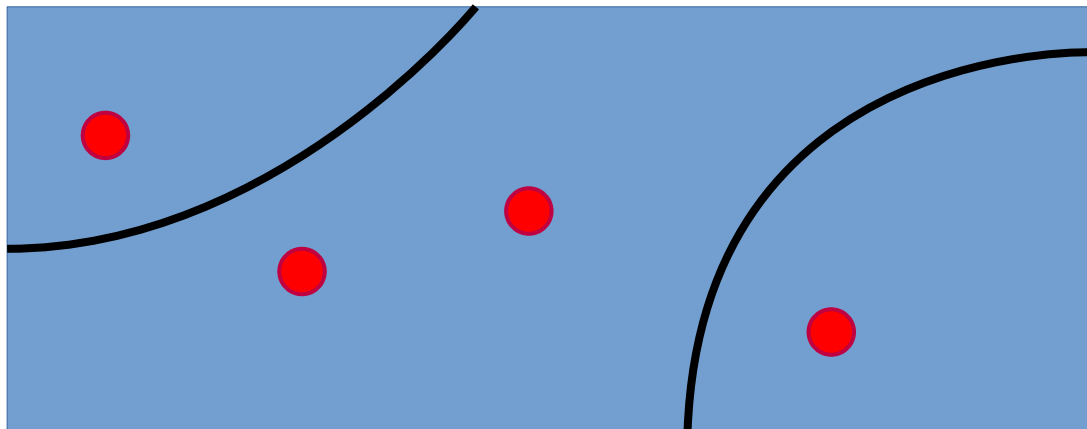
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You can already see how it relates to fuzzing & symbolic execution!

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- The emphasis is on defining (1) the scenario & (2) the oracle
- And we can use test generation strategies that we have already seen!
 - random testing
 - symbolic execution

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 - Test case reduction was also automatically applied
 - Now includes symbolic execution as a means of generation

Property Based Testing

- Traditional testing can be seen as example based

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- Property testing focuses on generic properties that should hold

$$\forall x : x > 0$$

$$\forall x, y, z : \varphi(x, y, z) \rightarrow \psi(x, y, z)$$

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 - Property testing tools can provide libraries to help define input spaces
 - Some domains may require substantial initial effort (similar to fuzzing)
- Because the process is so specification focused, it can also help developers understand the intent of their own code

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 - Idempotence `qsort(qsort(x)) == qsort(x)`
 - Invariants `qsort(x).size() == x.size()`

Digging Deeper

- What are good properties to check for a sorting function?

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- What if we have a sort over only one field?
- The actual properties to check can be more subtle than they appear!

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 - What happens with property based tests?
- Failing test cases even have test case reduction applied

In Practice: Hypothesis

- Hypothesis [<https://hypothesis.works/>]
 - Python, Java, ...
 - Presently uses random testing
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from hypothesis import given
from hypothesis.strategies import text

@given(text())
@example('')
def test_decode_inverts_encode(s):
    assert decode(encode(s)) == s
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```
@composite
def distinct_strings_with_common_characters(draw):
    x = draw(text(), min_size=1)
    y = draw(text(alphabet=x))
    assume(x != y)
    return (x, y)
```

- A rich set of primitives is available for more complex generator needs

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 - API is compatible with GoogleTest
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```
TEST(PrimePolynomial, OnlyGeneratesPrimes_NoStreaming) {
    symbolic_unsigned x, y, z;
    DeepState_Assume(x > 0);
    unsigned poly = (x * x) + x + 41;
    DeepState_Assume(y > 1);
    DeepState_Assume(z > 1);
    DeepState_Assume(y < poly);
    DeepState_Assume(z < poly);
    DeepState_Assert(poly != (y * z));
    DeepState_Assert(IsPrime(Pump(poly)));
}
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- By focusing more on goals rather than examples, it can have benefits even outside of testing
- Adoption can still require effort in defining good generators