



AUTOMATING TARGETED PROPERTY-BASED TESTING

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Property-Based Testing

- High-level, semi-automatic, black-box testing technique.
- Testing user-specified properties of the SUT.
- Examples:
 - QuickCheck (Haskell)
 - ScalaCheck (Scala)
 - PropEr (Erlang)
 - ...



PropEr

A QuickCheck-Inspired Property-Based Testing Tool for Erlang

Random Property-Based Testing

- PBT tool provides:
 - Random generators for basic types.
 - Language to write more complex generators.
- PBT tool automatically tests these properties:
 - Generates wide range of random inputs.
 - Runs the SUT with these inputs.
 - Checks if the properties hold.

Random Property-Based Testing

Generator

```
prop_list_reverse() ->  
  ?FORALL(L, list(integer()),  
    lists:reverse(lists:reverse(L)) == L).
```

Property should hold for all **L**

Property

Random Property-Based Testing

```
L=[]  
L=[2]  
L=[-5,-1,-8,1]  
L=[16,3,-23]  
L=[38,29,-28,12,-11,-3,-28,-6,9,-16,4,4]  
...
```

```
1> proper:quickcheck(example:prop_list_reverse(), 1000).  
..... 1000 dots .....  
OK: Passed 1000 test(s).
```

PBT of Sensor Networks

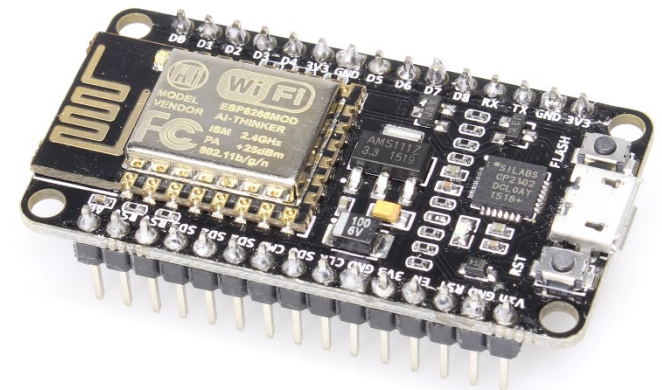
Setup:

- Sensor network
- Random distribution of UDP server and client nodes
- Client node periodically sends messages to server node

Test:

- Has X-MAC for any network a duty-cycle $> 25\%$?

(duty-cycle ::= % time the radio is on)



User-defined Generators

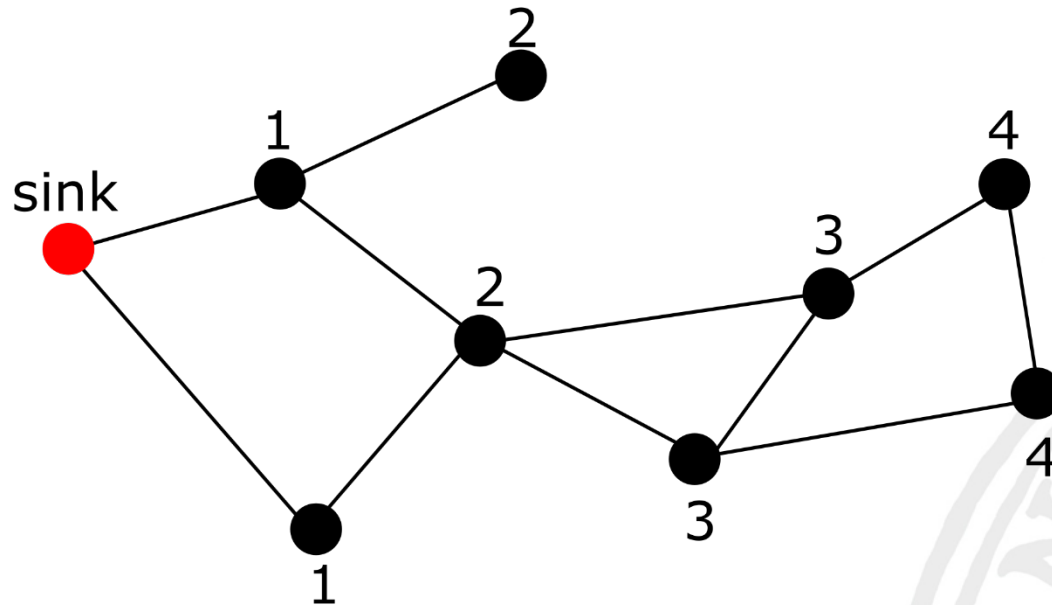
A generator for random graphs with **N** nodes:

```
graph(N) ->  
  Vs = lists:seq(1, N),  
  ?LET(Es, list(edge(Vs)), {Vs, lists:usort(Es)}).
```

```
edge(Vs) ->  
  ?SUCHTHAT({V1, V2}, {oneof(Vs), oneof(Vs)},  
            V1 < V2).
```

Great: We can easily generate random sensor networks!

Distance from Sink



On this graph, the maximum distance to sink is 4

Is there a large network with N nodes where the maximum distance to the sink $> N/2$?

Distance from Sink

```
prop_max_distance(N) ->  
  ?FORALL(G, graph(N),  
    begin  
      D = lists:max(distance_from_sink(G)),  
      D < (N div 2)  
    end).
```

```
2> proper:quickcheck(demo:prop_max_distance(42)).  
..... 100 dots .....  
OK: Passed 100 tests  
true  
3> proper:quickcheck(demo:prop_max_distance(42), 100000).  
..... 100000 dots .....  
OK: Passed 100000 tests  
true
```

Possible Solutions

- Write more involved (custom) generators?
- Guide the input generation:
use a search strategy, and
introduce a feedback-loop in the testing.

Targeted Property-Based Testing

- Combines Search Techniques with Property-Based Testing.
- Automatically guides input generation towards inputs with high probability of failing.
- Gather information during test execution in the form of **utility values (UVs)**.
- UVs capture how close input came to falsifying a property.

Targeted Property-Based Testing

Utility Value →

```
prop_max_distance(N) ->  
  ?FORALL_TARGETED(G, graph(N),  
    begin  
      D = lists:max(distance_from_sink(G)),  
      ?MAXIMIZE(D),  
      D < (N div 2)  
    end).
```

Now `prop_max_distance(42)` fails after 1,548 tests (on average).

Targeted Property-Based Testing

Simulated Annealing requires a Neighborhood Function (NF)

$nf(Base, Temperature)$

- Returns a neighbor (similar value) to a given *Base* value
- Neighbor distance can be scaled by the *Temperature*

Hand-written NF

```
graph_next(G, _T) ->
  Size = graph_size(G),
  ?LET(NewSize, neighboring_integer(Size),
  ?LET(Additional, neighboring_integer(Size div 10),
    begin
      {Removals, Additions} =
        case NewSize < Size of
          true ->
            {Additional + (Size - NewSize), Additional};
          false ->
            {Additional, Additional + (NewSize - Size)}
        end,
      ?LET(G_Del, remove_n_edges(G, Removals),
        add_n_edges(G_Del, Additions))
    end) ) .
```

Hand-written NF

```
graph_size({_, E}) -> length(E).
```

```
neighboring_integer(Base) ->  
  Offset = trunc(0.05 * Base) + 1,  
  ?LET(X, proper_types:integer(Base - Offset, Base + Offset),  
    max(0, X)).
```

```
add_n_edges({V, E}, N) ->  
  ?LET(NewEdges, proper_types:vector(N, edge(V)),  
    {V, lists:usort(E ++ NewEdges)}).
```

```
remove_n_edges({V, E}, 0) -> {V, E};
```

```
remove_n_edges({V, []}, _) -> {V, []};
```

```
remove_n_edges({V, E}, N) ->
```

```
  ?LET(Edge, proper_types:oneof(E),
```

```
    ?LAZY(remove_n_edges({V, lists:delete(Edge, E)}, N - 1))).
```

Neighborhood Function

- Neighborhood functions are significantly harder to write than random generators
31 vs 5 lines of code
- Must preserve all constraints and invariants of the input
- Makes TPBT difficult to use

Automating Targeted Property-Based Testing

- Construct the neighborhood function automatically from a random generator:
 - Random generator is problem-specific.
- Idea:
 - Reenact the decisions of the random generator.
 - Instead of choosing variables randomly, we choose values in the neighborhood of the previously generated one.

Example: Edge Generator

Constraint

Tuple Generator

`edge(Vs) -> /`
`?SUCHTHAT ({V1, V2},`
`V1 < V2) .`
`{oneof(Vs), oneof(Vs)},`

`Vs = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]`

Base Value: `{4, 8}`

Rule - for each element:

- Generate a random new element
- Generate a neighbor, or
- Leave as it is

Example: Edge Generator

Tuple Generator

```
edge(Vs) ->  
  ?SUCHTHAT({V1, V2}, {oneof(Vs), oneof(Vs)},  
             V1 < V2).
```

Vs = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]

Base Value: {4, 8}

Leave as it is

Change to a neighbor

Example: Edge Generator

oneof Generator

```
edge(Vs) ->  
  ?SUCHTHAT({V1, V2}, {oneof(Vs), oneof(Vs)},  
            V1 < V2) .
```

$Vs = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]$

Base Value: 8 Neighbor: 5

Rule - exchange to a random element

Example: Edge Generator

Constraint

Tuple Generator

`edge(Vs) -> /`
`?SUCHTHAT ({V1, V2},`
`V1 < V2) .`
`{oneof(Vs), oneof(Vs)},`

`Vs = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]`

Base Value: {4, 8} Neighbor: {4, 5}

- Check constraint:
 - If fulfilled: done
 - Else: retry

Automating Targeted Property-Based Testing

- Rules for all built-in types of PropEr.
- More complex types (constructed with ?**LET**) require some additional techniques:
 - Matching
 - Caching
- It is possible to adjust the generation process by overwriting rules with own ones.

Limitations

- Certain nested generators with multiple **LET**
 - if the constraints for the inner generators depend on values generated by the outer generators.
- Recursive generators.

PBT of Sensor Networks

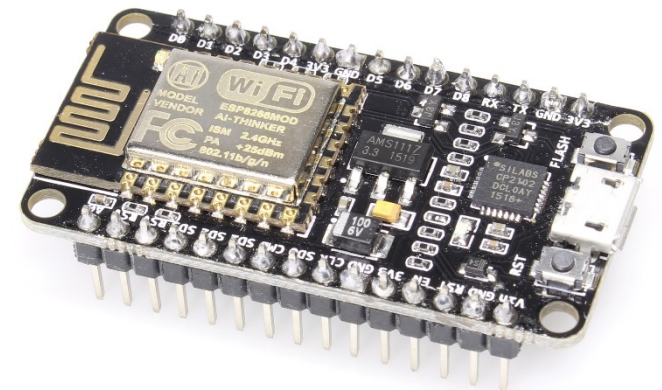
Setup:

- Sensor network.
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Test:

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(duty-cycle ::= % time the radio is on)



Case Study 1

Random PBT

- **Mean Time to Failure: 7h46m**

Targeted PBT with hand written NF (100 loc)

- **Mean Time to Failure : 2h12m**

Targeted PBT with constructed NF

- **Mean Time to Failure : 2h19m**

Case Study 2

$$\frac{i(pc) = \text{Noop}}{\boxed{pc} \mid \boxed{s} \mid \boxed{m} \Rightarrow \boxed{pc+1} \mid \boxed{s} \mid \boxed{m}} \quad (\text{NOOP})$$

$$\frac{i(pc) = \text{Push } v}{\boxed{pc} \mid \boxed{s} \mid \boxed{m} \Rightarrow \boxed{pc+1} \mid \boxed{v : s} \mid \boxed{m}} \quad (\text{PUSH})$$

$$\frac{i(pc) = \text{Pop}}{\boxed{pc} \mid \boxed{v : s} \mid \boxed{m} \Rightarrow \boxed{pc+1} \mid \boxed{s} \mid \boxed{m}} \quad (\text{POP})$$

- Definitions for an abstract machine.
- Test: Do these definitions fulfill a certain security criteria?
(Noninterference)

Cătălin Hrițcu, et al. "Testing noninterference, quickly." *Journal of Functional Programming*, 26 (2016).

Case Study 2

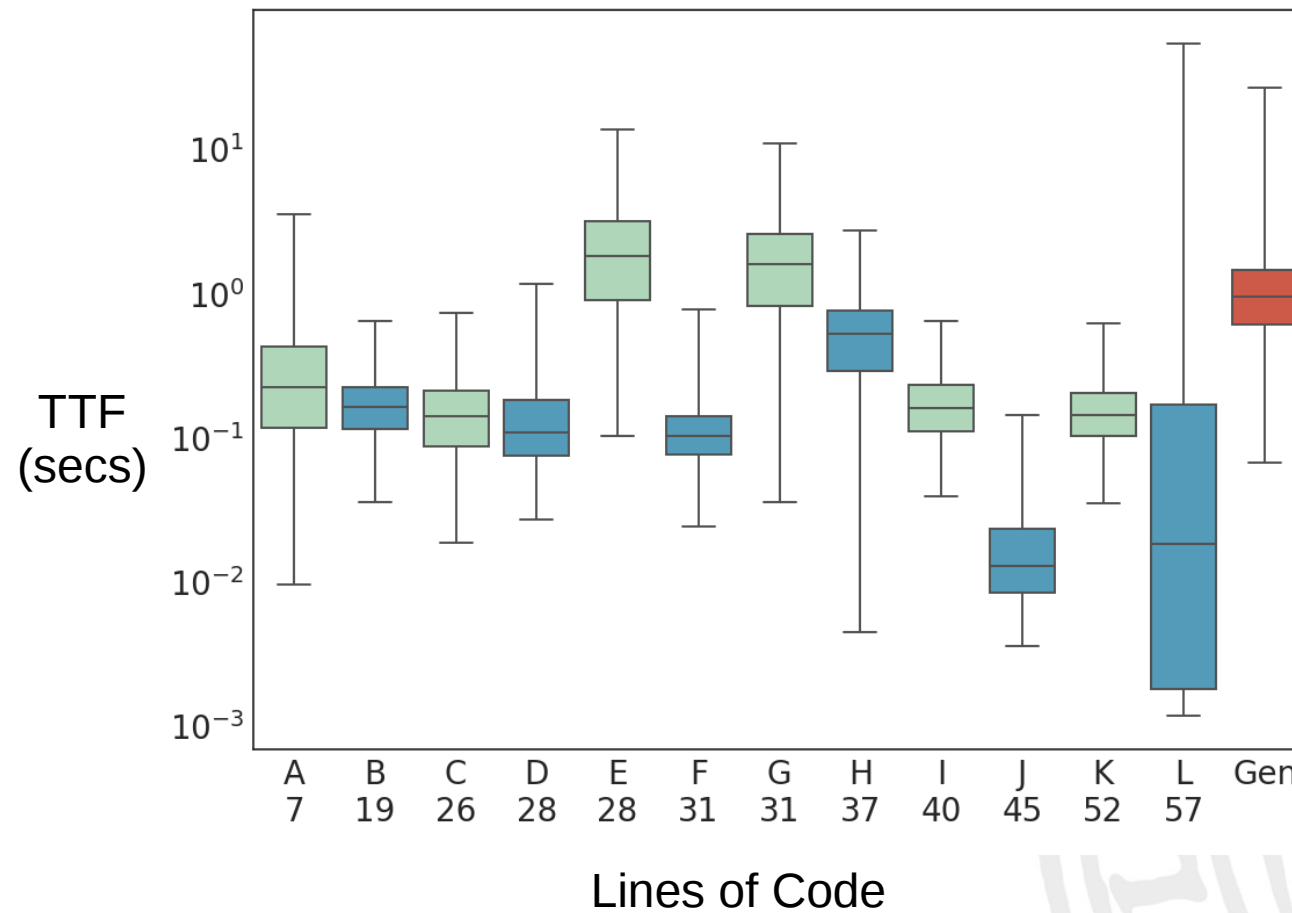
- **Random PBT - Sequence:** programs are a random list of instructions chosen with a fine-tuned weighted distribution
- **Targeted PBT - List:** hand written NF for lists; list elements are either added new or removed
- **Targeted PBT - Constructed:** constructed NF from the Sequence generator

	Random PBT	Targeted PBT	Targeted PBT
	Sequence	List	Constructed
ADD	5800,57	271,68	489,93
LOAD	7764,15	341,30	447,25
STORE A	16997.81	2634,80	3685,32

User Study

- We asked students from an advanced functional programming course (M.Sc.) to program a NF for testing a targeted property.
- All students were familiar with random and targeted PBT.
- Compared the hand-written NFs to the constructed one.

User Study



Targeted Property-Based Testing

```
prop_max_distance(N) ->
  ?FORALL_TARGETED(G, graph(N)).
begin
  D = lists:max
  ?MAXIMIZE(D)
  D < (N div 2
end)).
```

Utility
Value



Now `prop_max_distance`
(on average).

Hand-written NF

```
graph_next(G, _T) ->
  Size = graph_size(G),
  ?LET(NewSize, neighboring integer(Size),
    _integer(Size div 10),
    =
    of
    Size - NewSize), Additional};
ditional + (NewSize - Size)}
edges(G, Removals),
l, Additions))
```

PropEr

A QuickCheck-Inspired Property-Based Testing Tool for Erlang

<http://proper.softlab.ntua.gr/>

Examp

```
edge(Vs) ->
  ?SUCHTHAT({V1, V2}, {one
    V1 < V2}).
```

Vs = [1, 2, 3, 4, 5, 6, 7

Base Value: {4, 8}

Leave as it is

Change to a neighbor



User Study

