

We've got you Covered: Type-Guided Repair of Incomplete Input Generators

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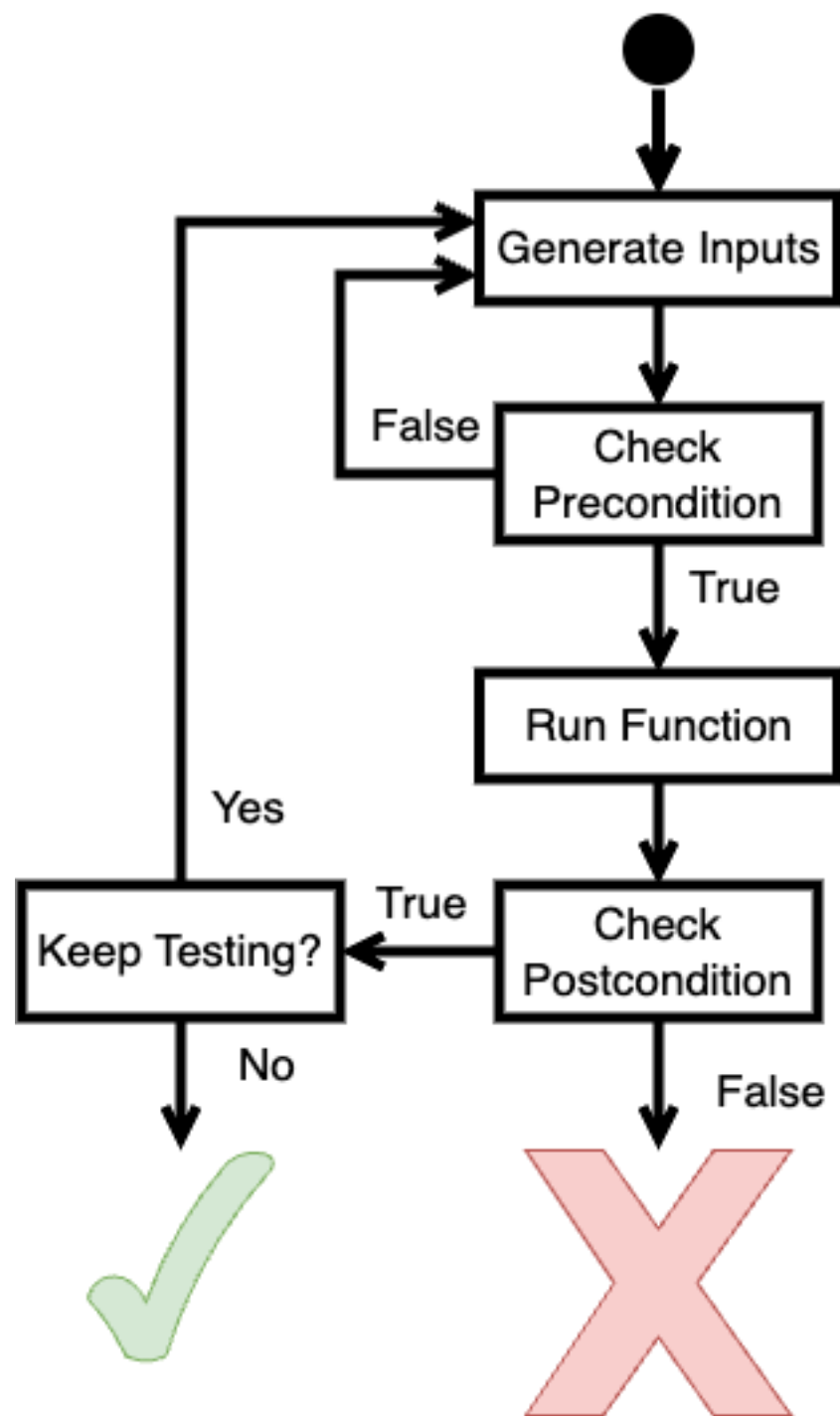
**Benjamin
Delaware**







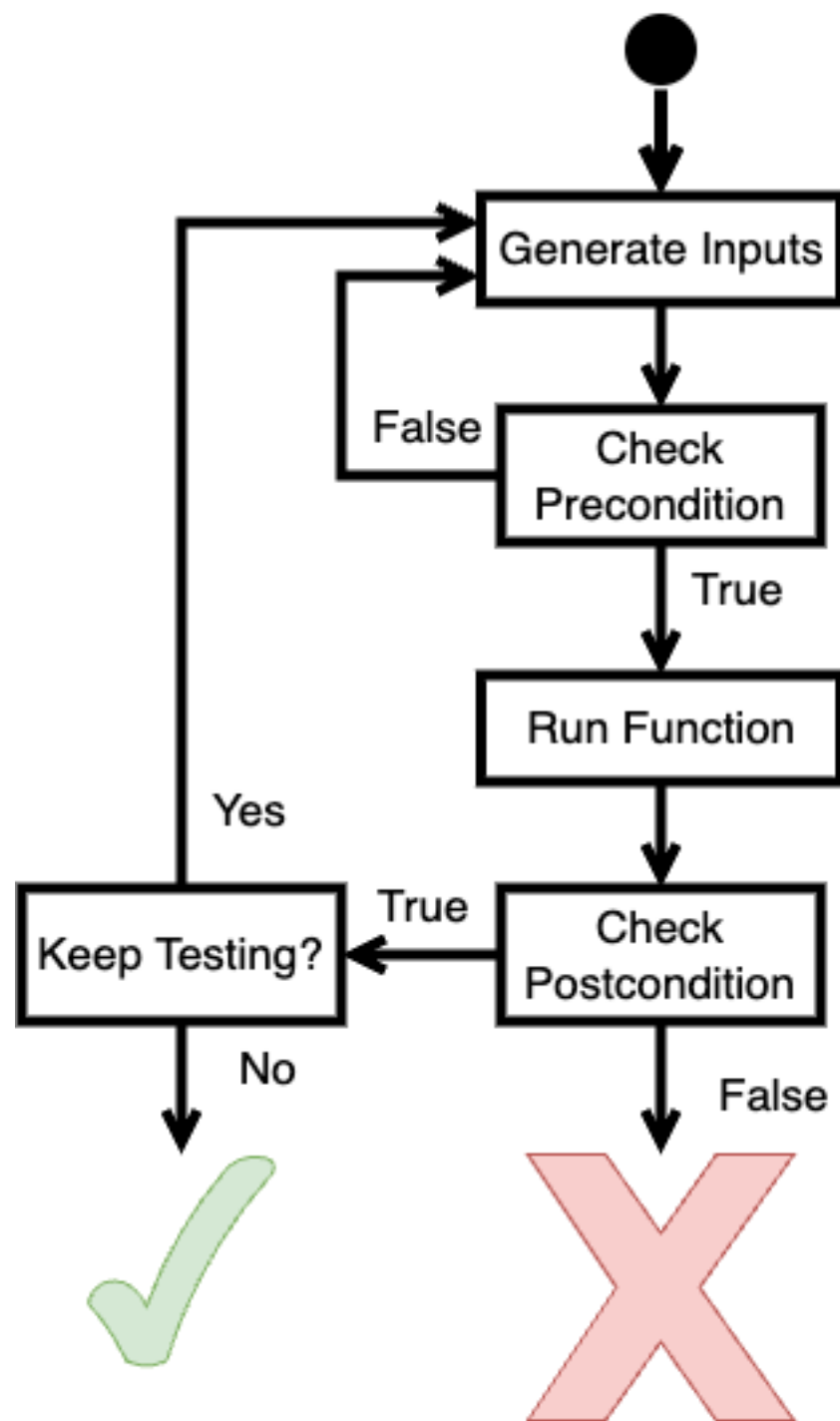
PBT in a Nutshell



Property-Based Testing is an **automated testing strategy**.

A variety of frameworks in all of your favorite languages!

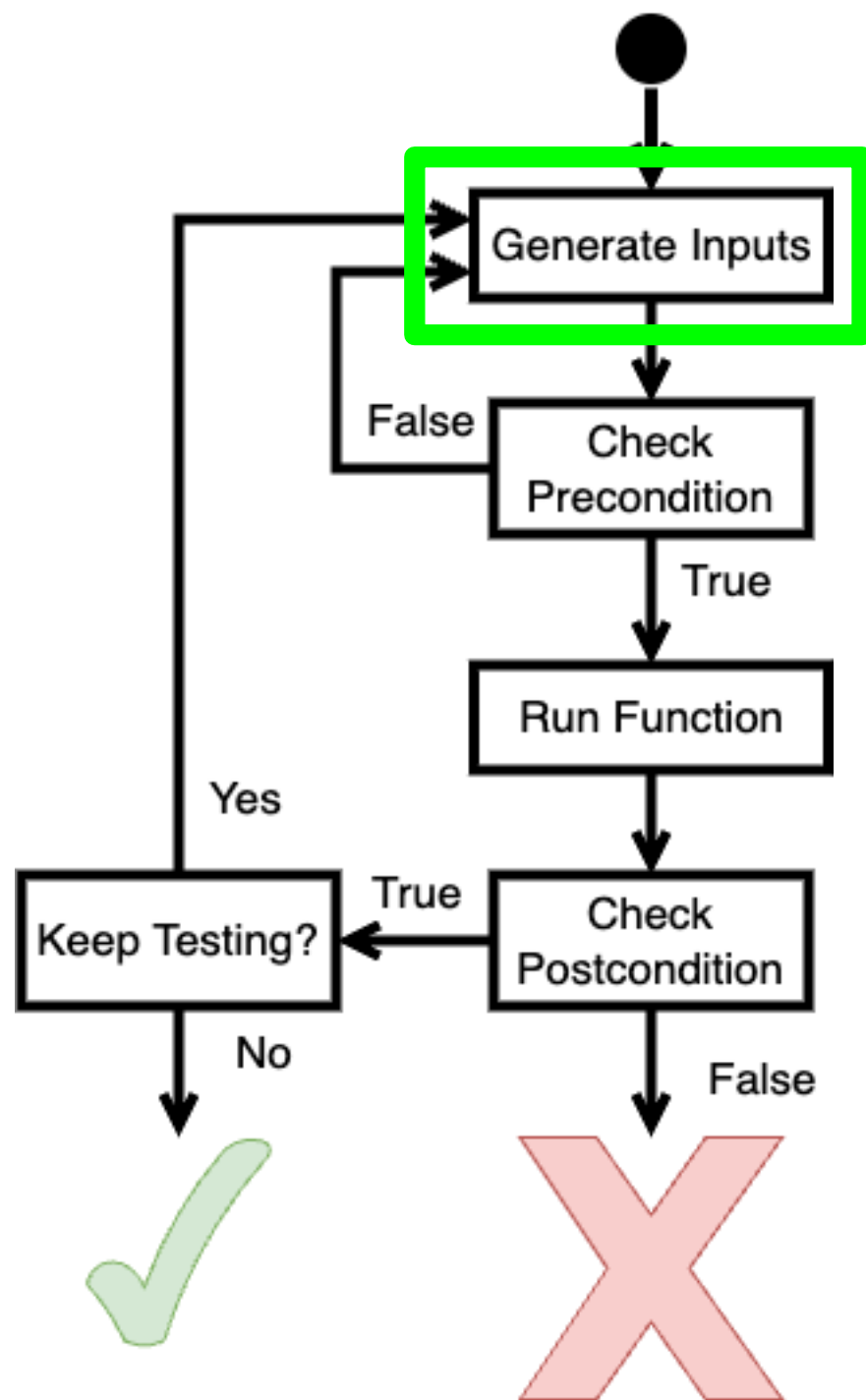
PBT in a Nutshell



Generate a bunch of random well-sized lists

```
let bad_test = Test.make
~count:20000
~name:"tests_R_us"
(default_gen ())
(fun (n, l) ->
  assume(List.length l <= n) ;
  func l n |> check_post)
```

PBT in a Nutshell



Built-in or are automatically derivable

```
[ @@ gen ] or # [ derive (Arbitrary) ]
```

```
let bad_test = Test.make
```

```
~count:20000
```

```
~name:"tests R us"
```

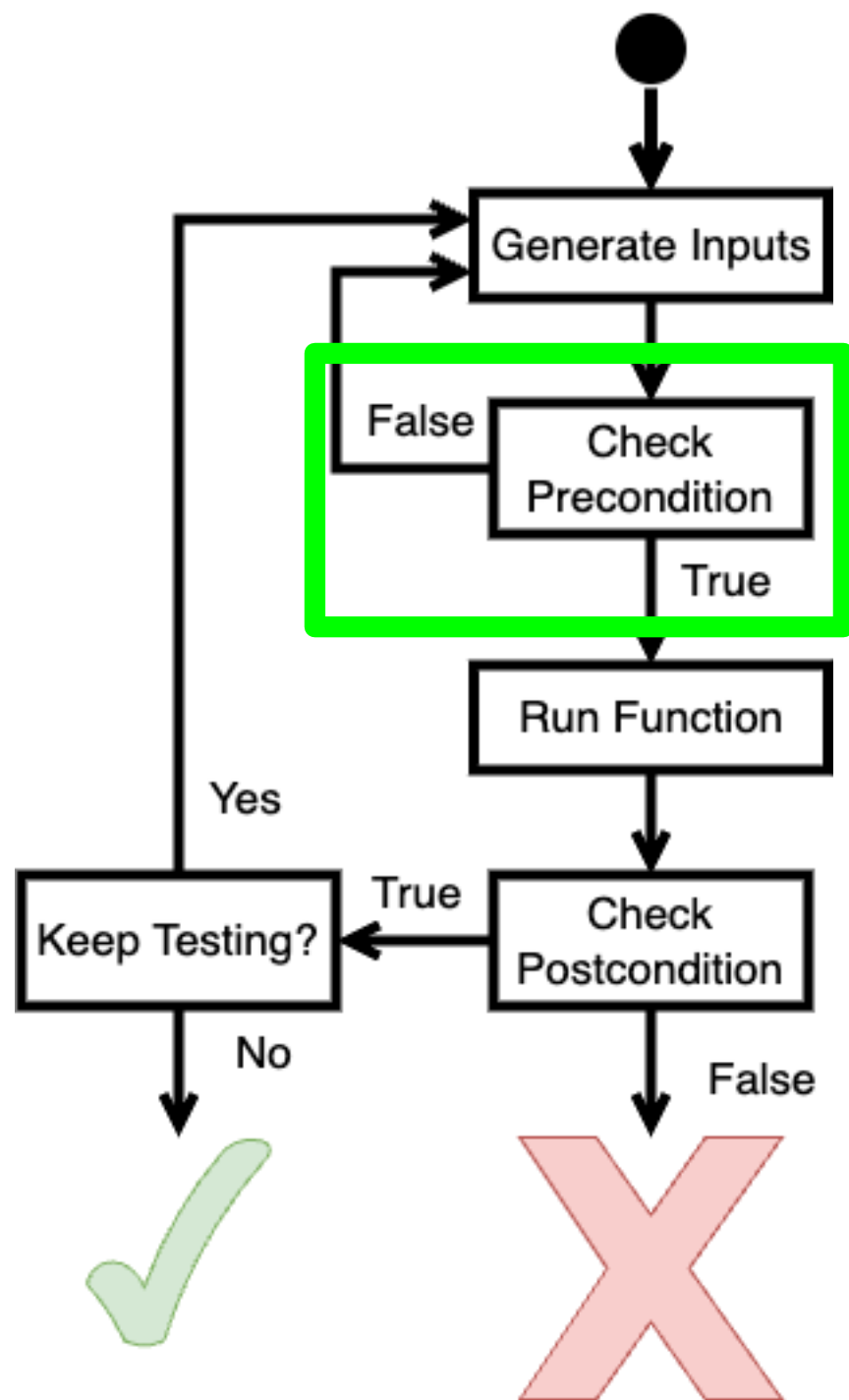
```
(default_gen ())
```

```
(fun (n, l) ->
```

```
  assume (List.length l <= n) ;
```

```
  func l n |> check_post)
```

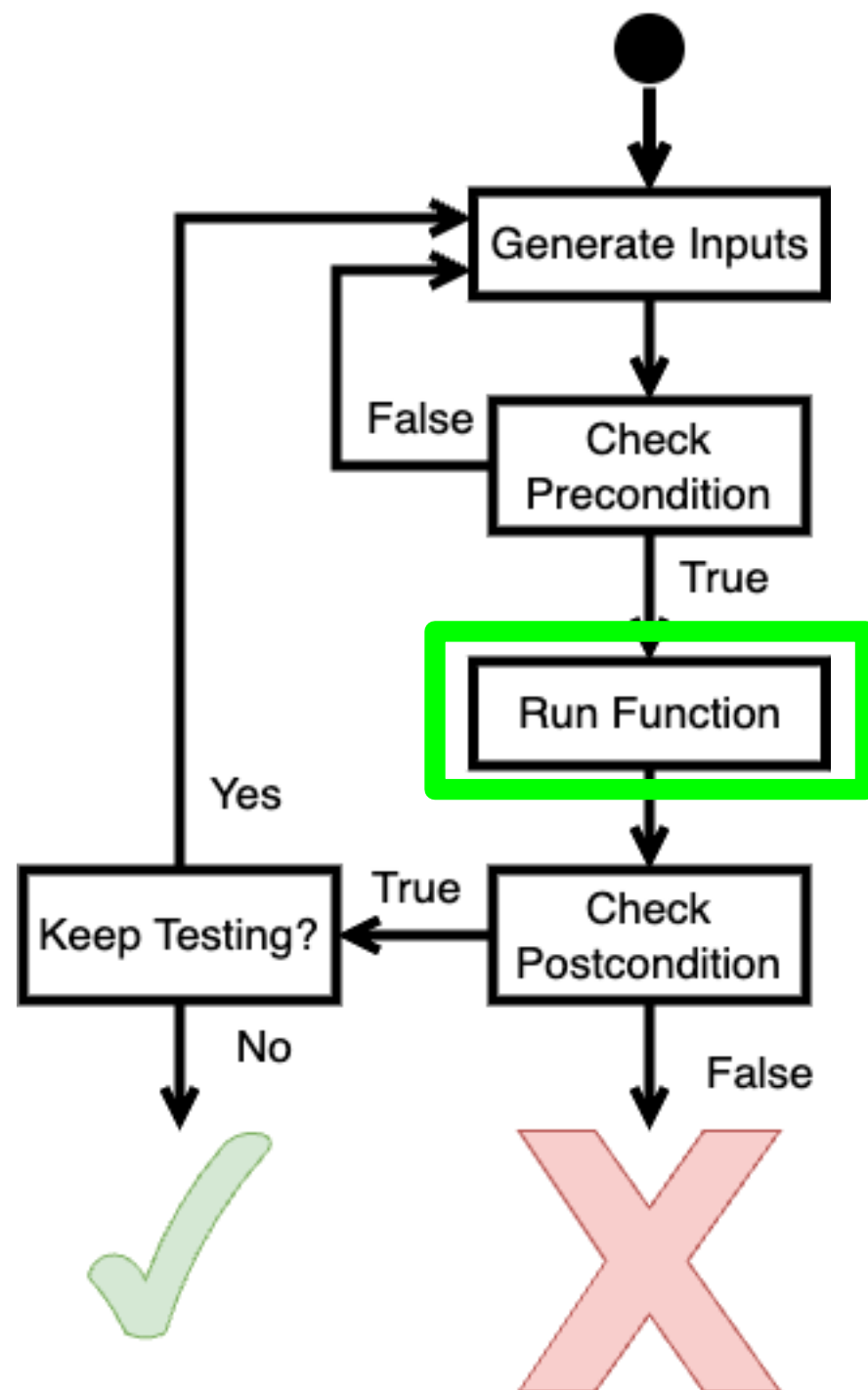
PBT in a Nutshell



Generated inputs must satisfy the precondition

```
let bad_test = Test.make
~count:20000
~name:"tests_R_us"
(default_gen ())
(fun (n, l) ->
  assume(List.length l <= n) ;
  func l n |> check_post)
```

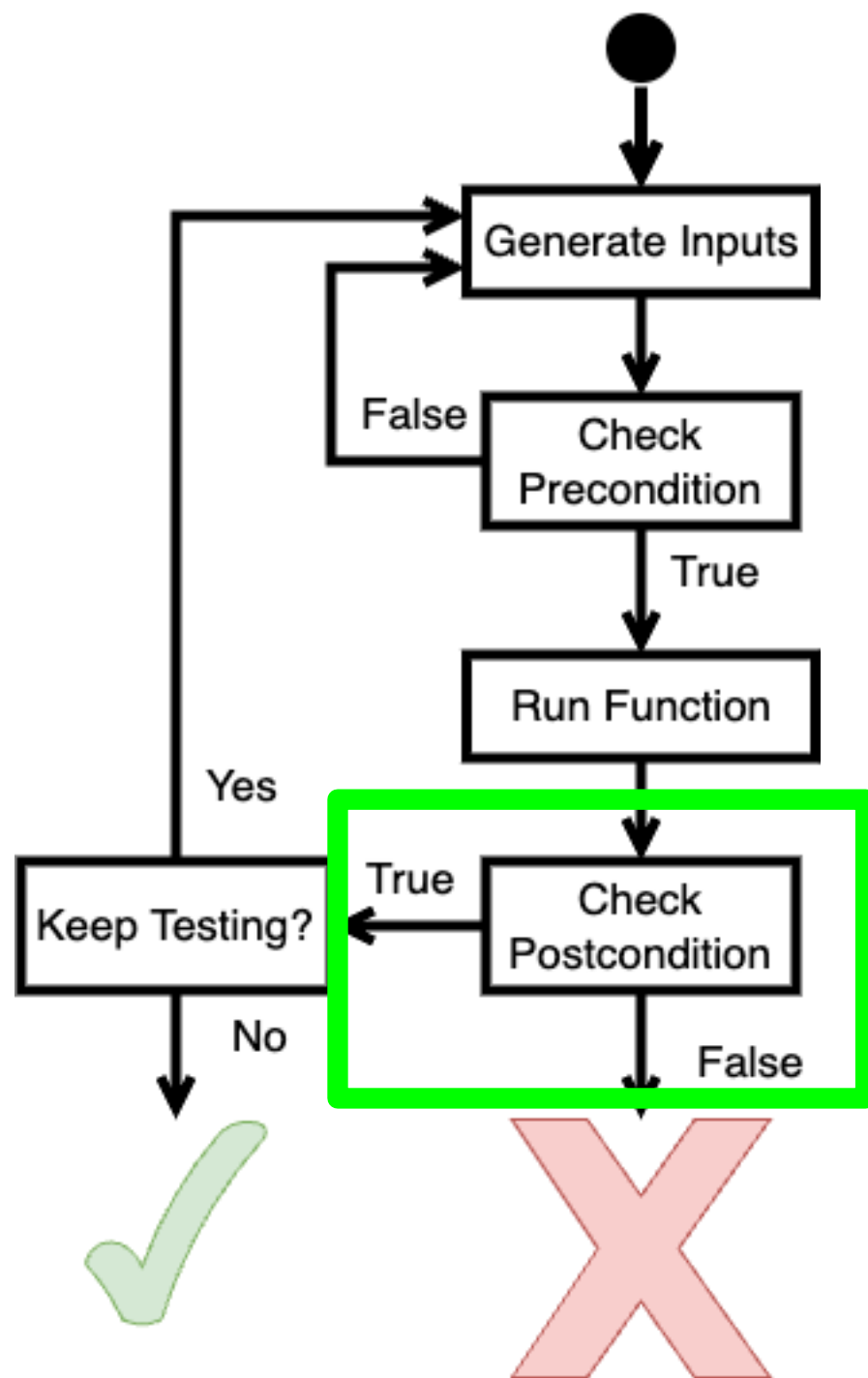
PBT in a Nutshell



User function under test

```
let bad_test = Test.make
~count:20000
~name:"tests_R_us"
(default_gen ())
(fun (n, l) ->
  assume(List.length l <= n) ;
  func l n |> check_post)
```

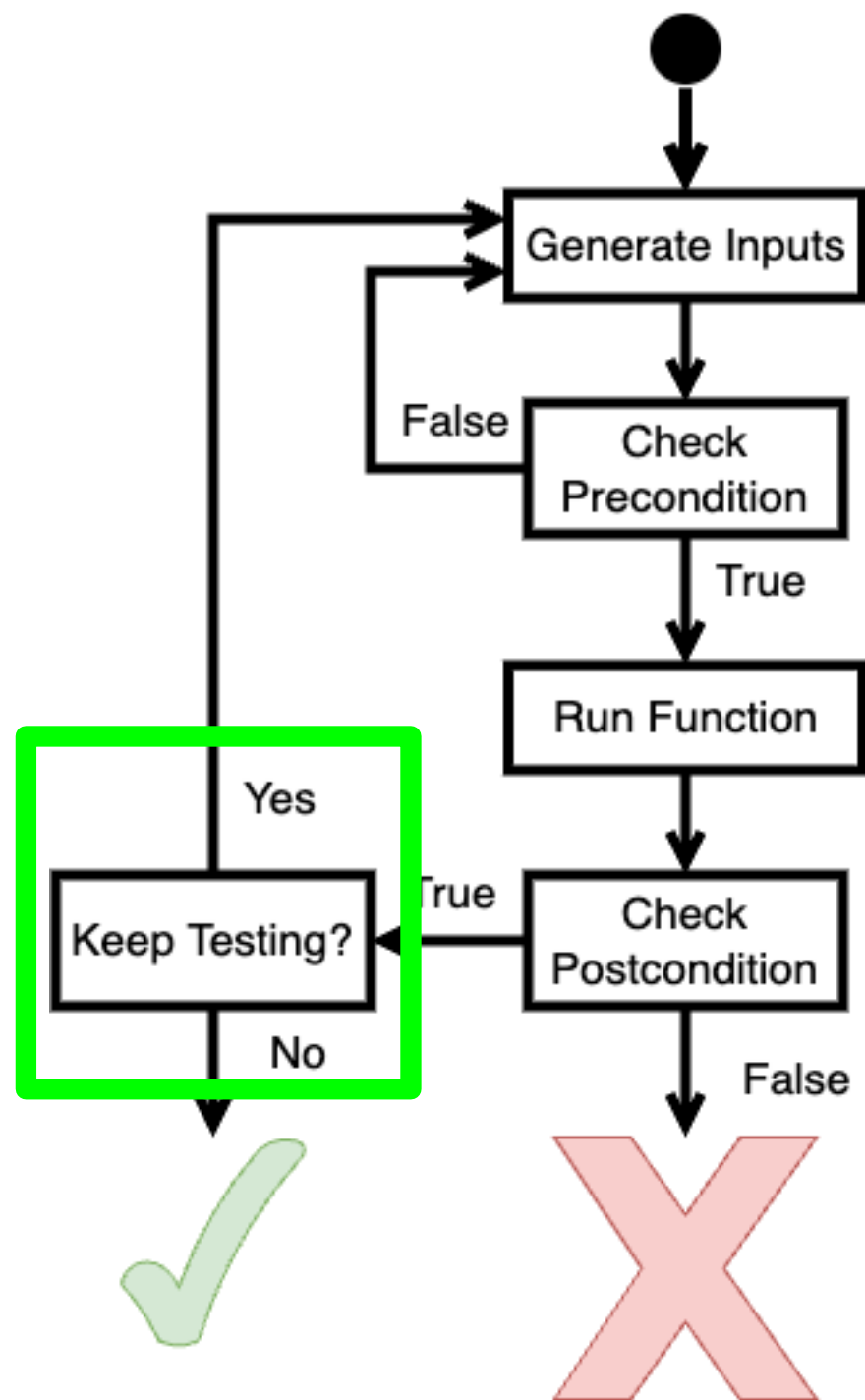
PBT in a Nutshell



User post-condition maps output to True/False

```
let bad_test = Test.make
~count:20000
~name:"tests_R_us"
(default_gen ())
(fun (n, l) ->
  assume(List.length l <= n) ;
  func l n |> check_post)
```

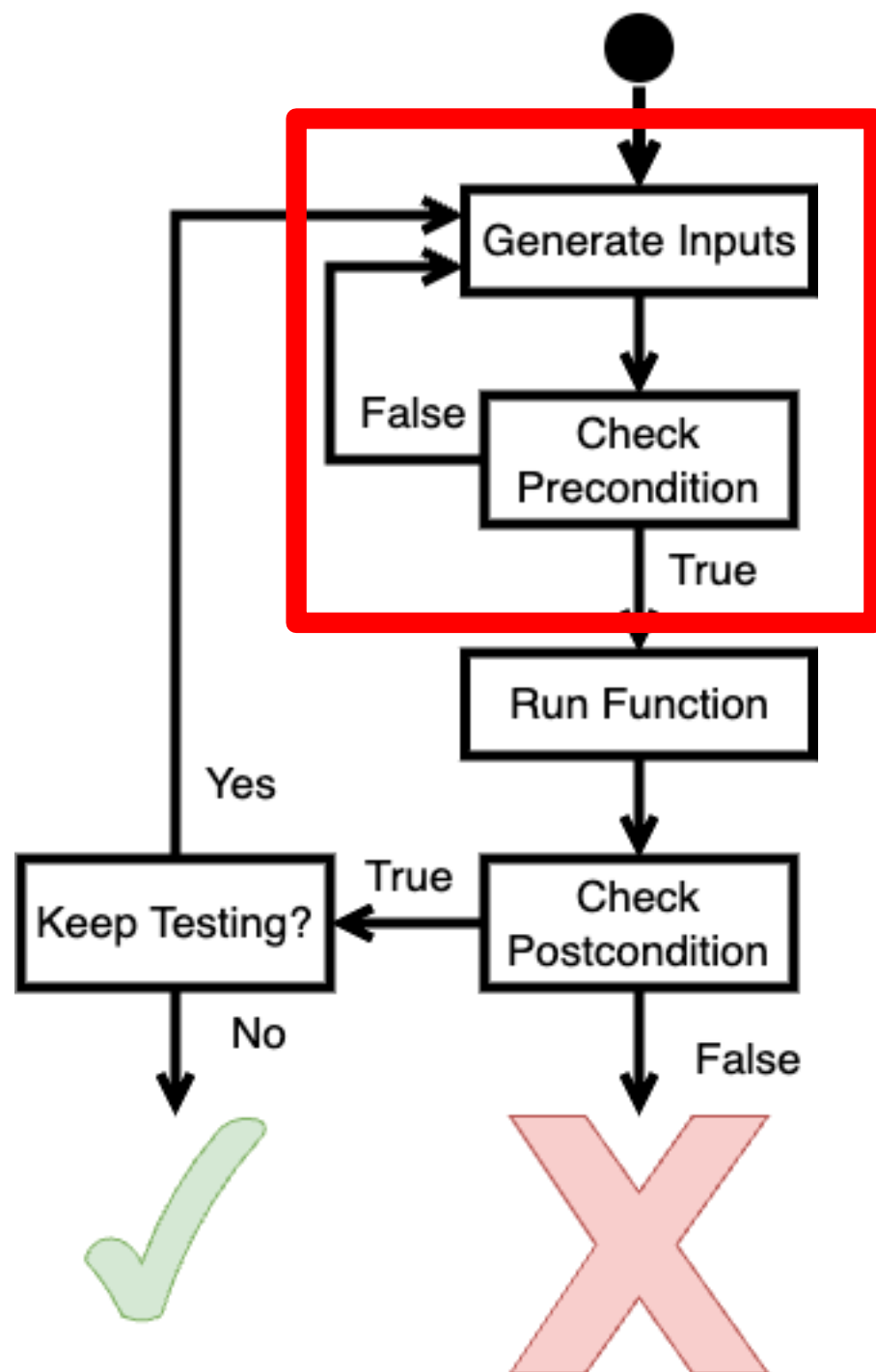
PBT in a Nutshell



Run for 20k attempts or until
postcondition fails

```
let bad test = Test.make  
~count:20000  
~name:"tests_R_us"  
(default_gen ())  
(fun (n, l) ->  
  assume(List.length l <= n) ;  
  func l n |> check_post)
```

PBT in a Nutshell



Focus of this talk

Wasted work where we aren't testing the user's function

So you think you can test?

How well does the default generator do?

```
let bad_test = QCheck.(Test.make
  ~count:20000
  ~name:"tests R us"
  (default_gen ())
  (fun (n, l) ->
    assume (List.length l <= n);
    func l n |> check_post))
```

So you think you can test?

Sized List: only about 50% of inputs generated inputs are valid

```
> dune exec example
generated error fail pass / total      time test name
[✓] 20200      0    0 10050 / 20000      0.3s tests_R_us
=====
success (ran 1 tests)
```

```
assume (List.length l <= n);
fun l n |> check_post)
```

So you think you can test?

More restrictive precondition -> more rejected inputs

```
let bad_test = QCheck.(Test.make
  ~count:20000
  ~name:"tests_R_us"
  (default_gen ())
  (fun () ->
    assume (is_even_list 1);
    fun l n |> check_post))
```

So you think you can test?

More restrictive precondition -> more rejected inputs

Even List: About 10% of inputs are valid

```
> dune exec example
generated error fail pass / total time test name
[✓] 20200      0    0 2138 / 20000 0.3s tests_R_us
=====
success (ran 1 tests)
```

```
assume (is_even_list l);
fun l n |> check_post)
```

So you think you can test?

Specialized generators -> all inputs valid

```
let bad_test = QCheck.(Test.make
  ~count:20000
  ~name:"tests_R_us"
  (even_list_gen ())
  (fun (l) ->
    assume (is_even_list l);
    func l n |> check_post))
```

So you think you can test?

Specialized generators -> all inputs valid

```
> dune exec example
generated error  fail  pass / total  time test name
[✓] 20000      0      0 20000 / 20000  0.2s tests_R_us
=====
success (ran 1 tests)
```

```
(fun (l) ->
```

```
  assume (is_even_list l);
  func l n |> check_post))
```

So you think you can test?

Users need to write **specialized** generators to ensure enough **valid** inputs

```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then [ 2 * int_gen () ]  
  else  
    if bool_gen () then [ 2 * int_gen () ]  
    else  
      2 * int_gen () :: even_list_gen (s - 1)
```

So you think you can test?

Users need to write **specialized** generators to ensure enough **valid** inputs

```
let rec even_list_gen (s : int) : int list =  
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```

Different precondition? Different generator needed





OB5: *Developers see writing generators as a distraction, preferring to use derived generators.* Since PBT is often done in the midst of development, developers are reluctant to slow down and write a generator; the task was seen as both difficult and time-consuming. Instead, as seen in §4.4, many participants opted to test with generators derived from the types of their programs.

Goldstein et al., "Property-Based Testing in Practice," ICSE 2024

Program Synthesis to the Rescue?

High Leverage, Low Investment Opportunity for Program Synthesis



Program Synthesis to the Rescue?

Program synthesis requires a **specification** of the users intent

even_list_gen: “**Only** non-empty lists of even ints”

Program Synthesis to the Rescue?

Program synthesis over an initial **template** with components

even_list_gen: “**Only** non-empty lists of even ints”

+

```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then Err  
  else Err
```

Program Synthesis to the Rescue?

Many **valid** programs make for **disappointing** generators

even_list_gen: “**Only** non-empty lists of even ints”



```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then Err  
  else Err
```



?

Always producing [0] is valid

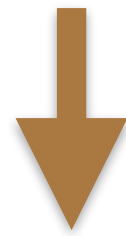
```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then [ 0 ]  
  else [ 0 ]
```

Program Synthesis to the Rescue?

even_list_gen: "**Every** non-empty lists of even ints"



```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then Err  
  else Err
```



Good generators produce **every**
valid input

```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then [ 2 * int_gen () ]  
  else  
    if bool_gen () then [ 2 * int_gen () ]  
    else  
      2 * int_gen () :: even_list_gen (s - 1)
```

Missing ingredient: “Coverage”

even_list_gen: “**Every** non-empty lists of even ints”

```
even_list_gen: {s:int | v >= 0}  
-> [int list | all_evens(v) && 0<len(v)<=s+1]
```

“Coverage Types”



Formally:

“**only**” -> safety property

“**every**” -> coverage property

Covering All the Bases: Type-Based Verification of Test Input Generators

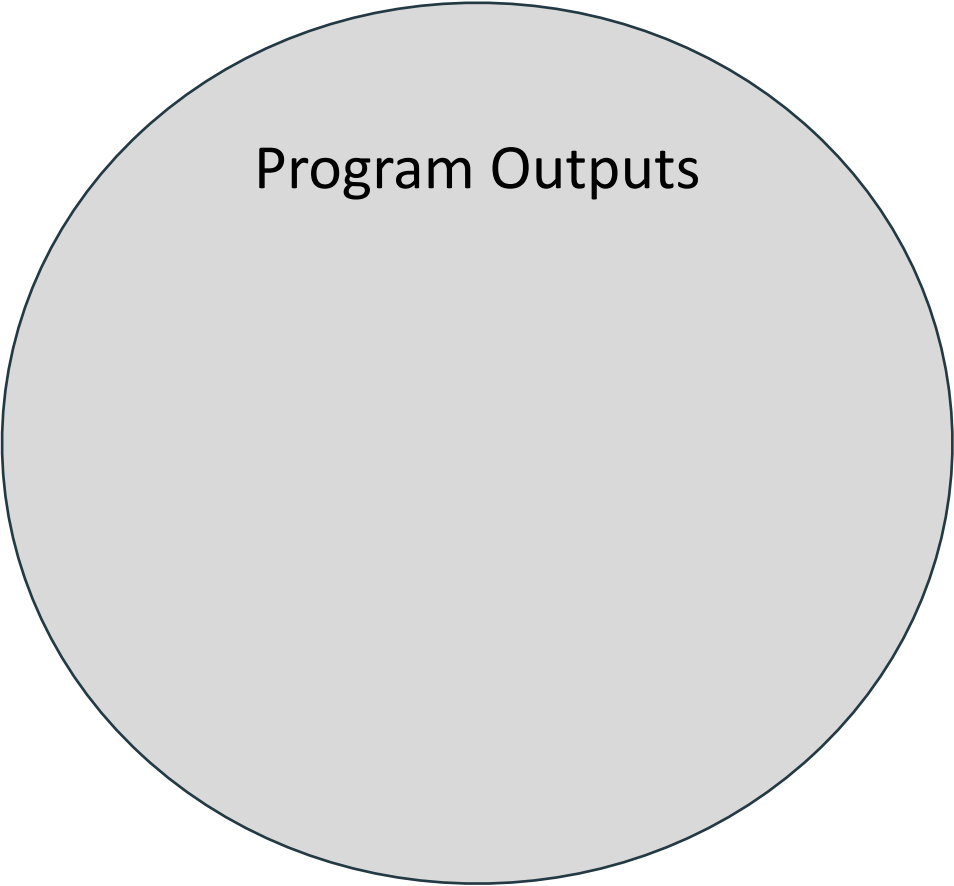
ZHE ZHOU, Purdue University, USA

ASHISH MISHRA, Purdue University, USA

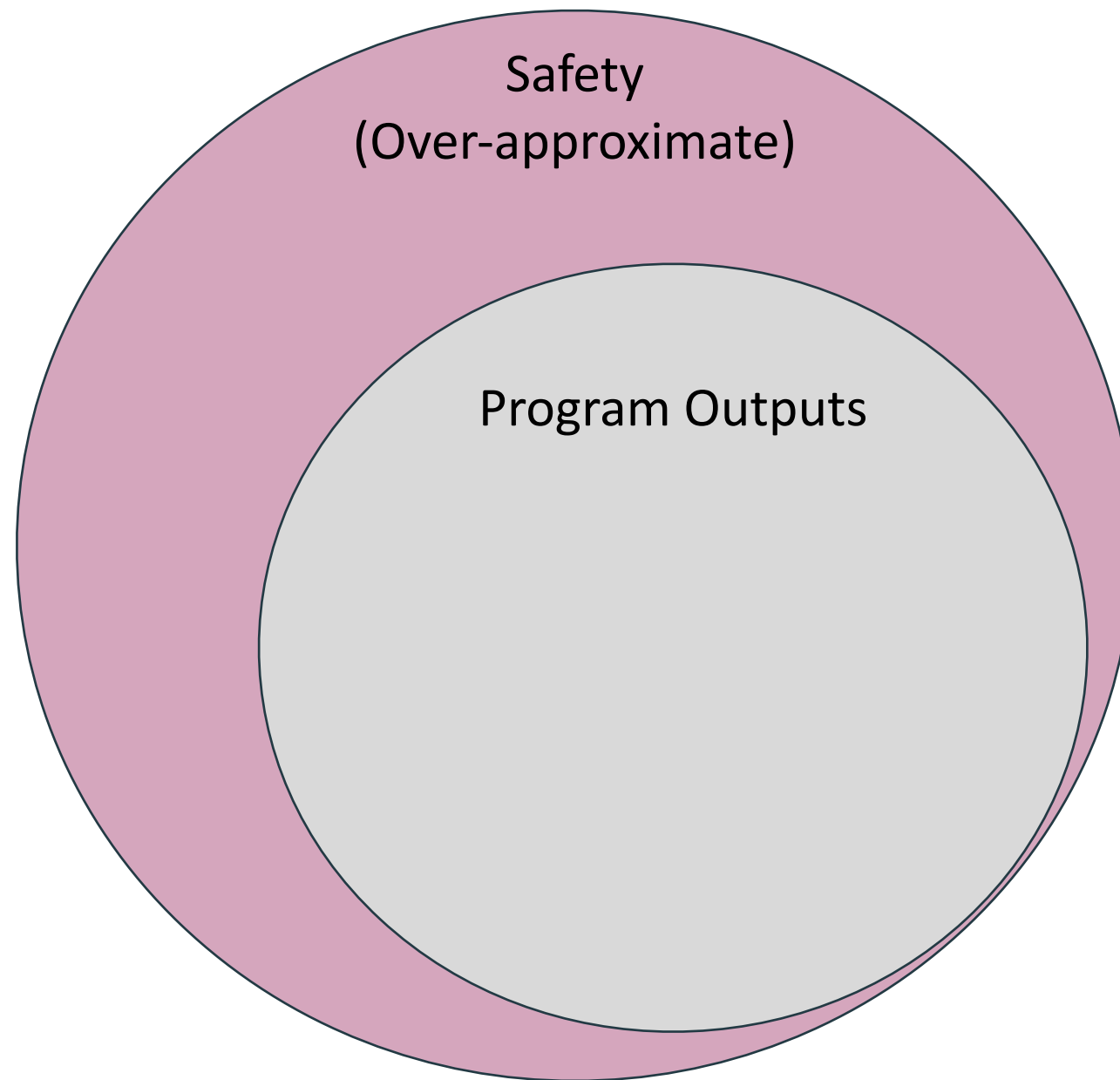
BENJAMIN DELAWARE, Purdue University, USA

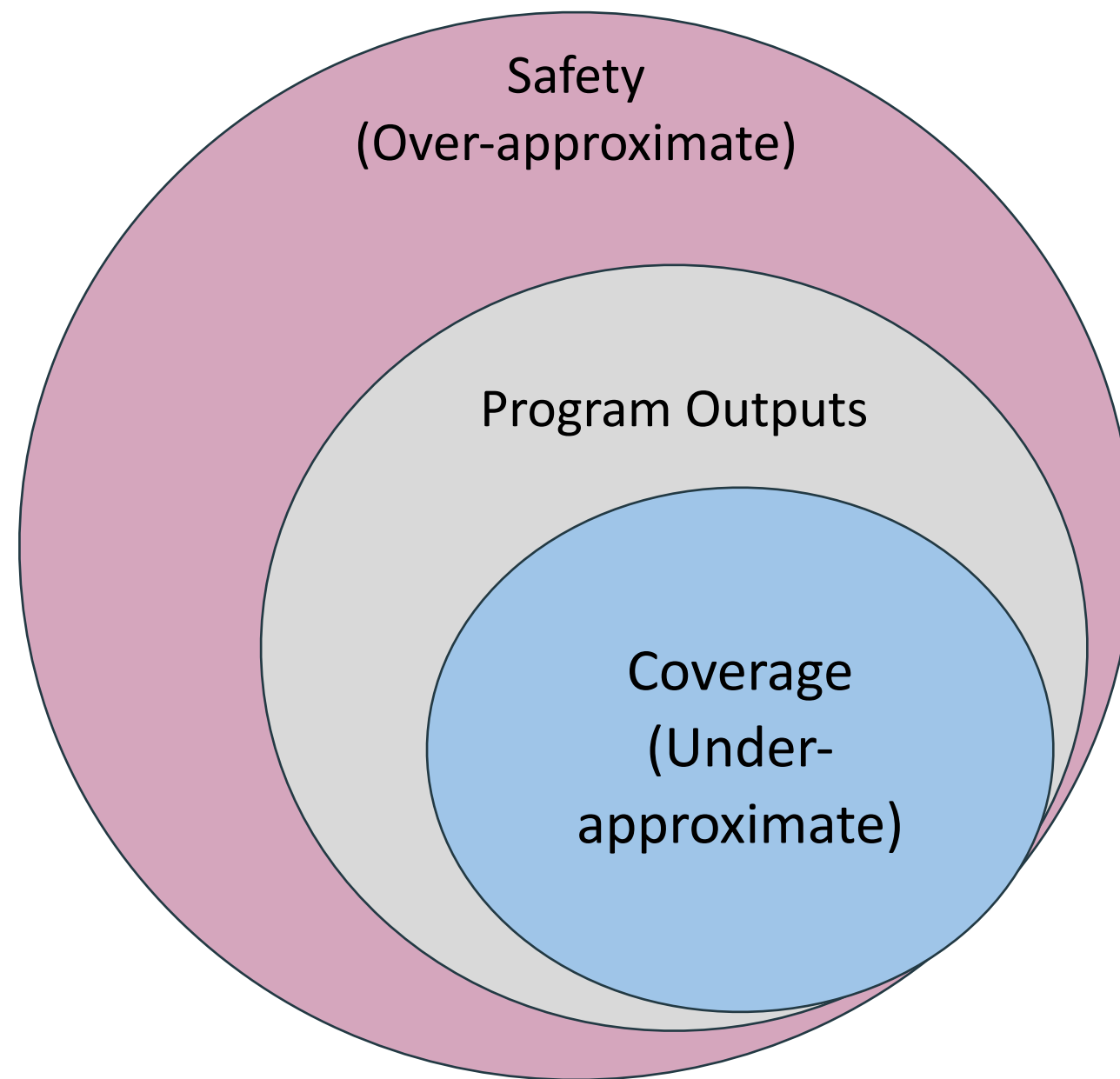
SURESH JAGANNATHAN, Purdue University, USA

Test input generators are an important part of property-based testing (PBT) frameworks. Because PBT is intended to test deep semantic and structural properties of a program, the outputs produced by these generators can be complex data structures, constrained to satisfy properties the developer believes is most relevant to testing the function of interest. An important feature expected of these generators is that they be capable of producing *all* acceptable elements that satisfy the function’s input type and generator-provided constraints. However, it is not readily apparent how we might validate whether a particular generator’s output satisfies this *coverage* requirement. Typically, developers must rely on manual inspection and post-mortem analysis of test runs to determine if the generator is providing sufficient coverage; these approaches are error-prone and difficult to scale as generators become more complex. To address this important concern, we present a new refinement type-based verification procedure for validating the coverage provided by input test generators, based on a novel interpretation of types that embeds “*must-style*” underapproximate reasoning principles as a



Program Outputs



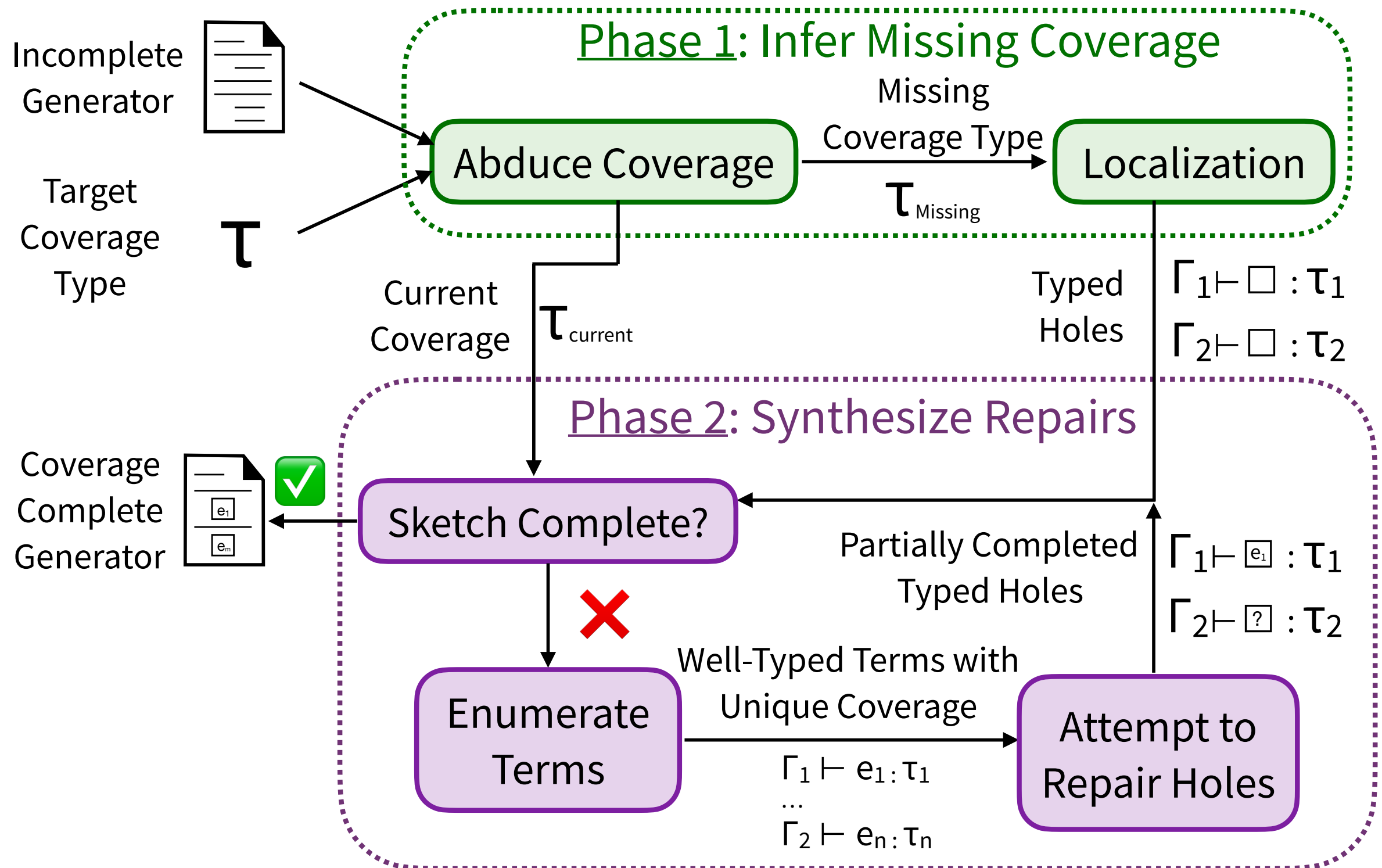


Coverage + Synthesis = Cobb

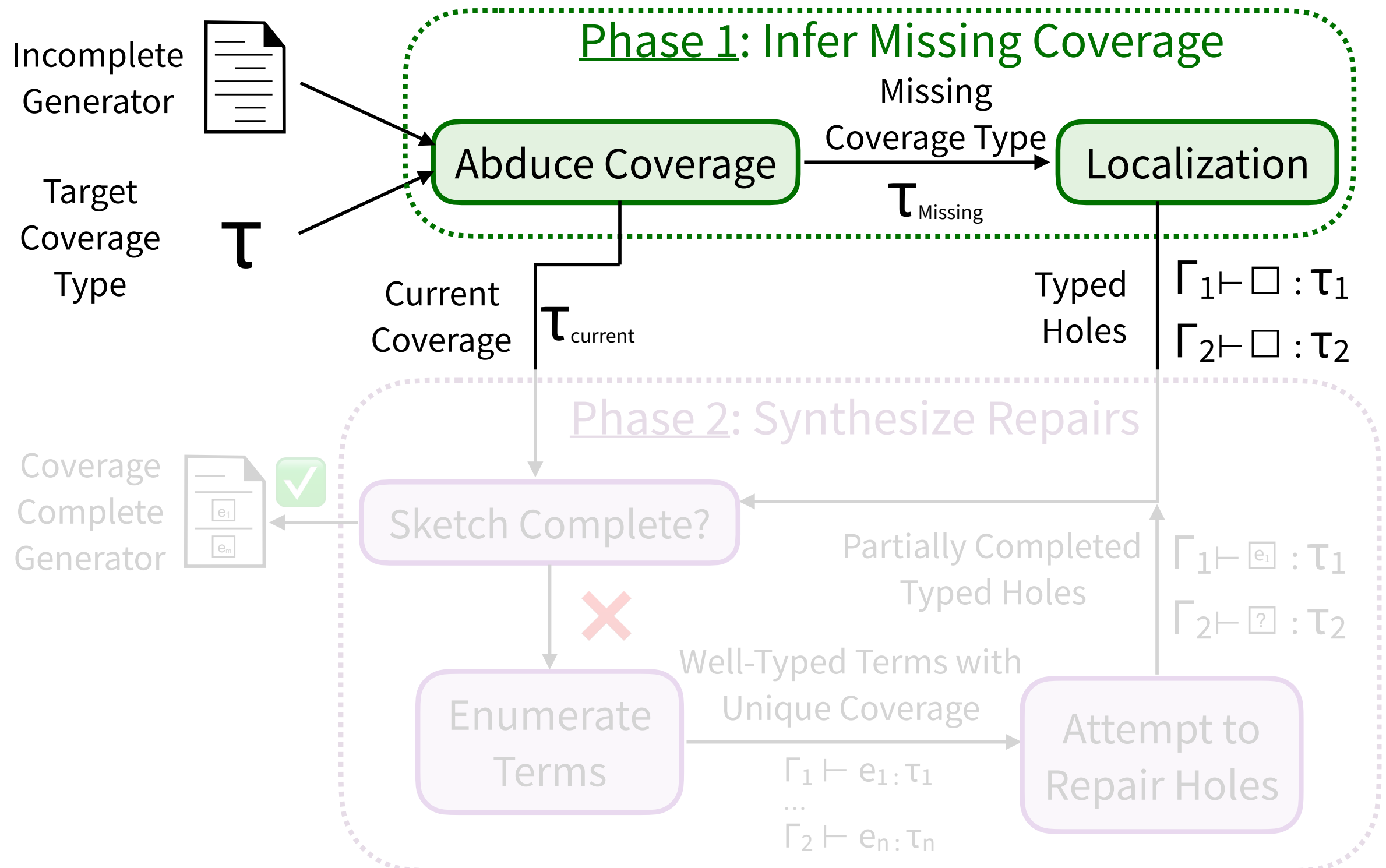
	Safety	Coverage
Type-Based Verification	Liquid Haskell Flux RefinedRust RefinedC refined(scala) ...	Poirot
Type-Based Synthesis	Synquid Cobalt ReSyn SuSLik ...	Cobb (This work)

So how does Cobb work?

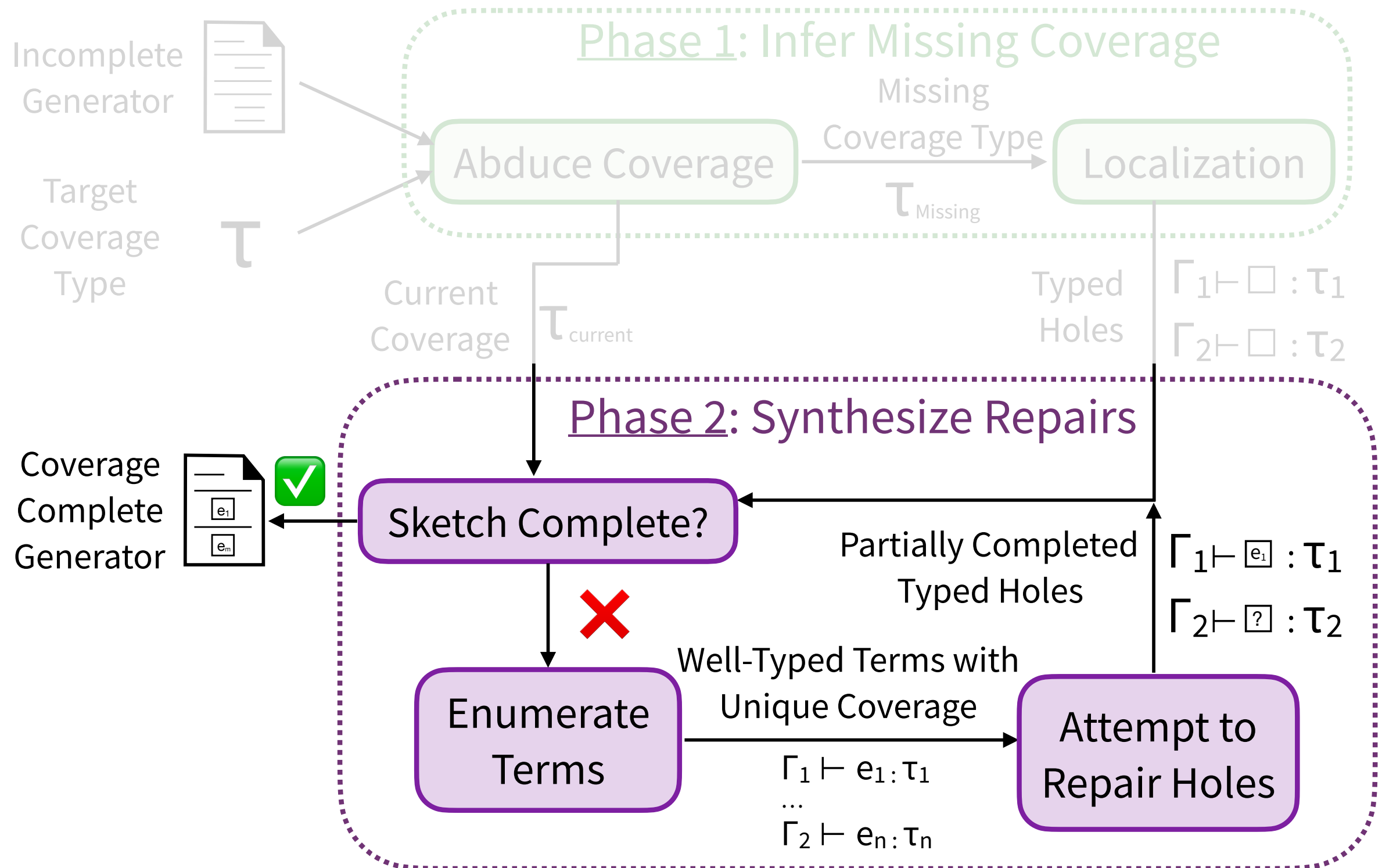
Cobb in a Nutshell



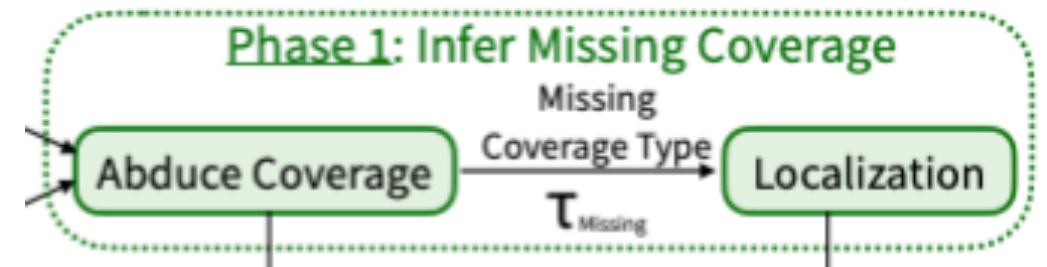
Cobb in a Nutshell



Cobb in a Nutshell



Setting up Synthesis



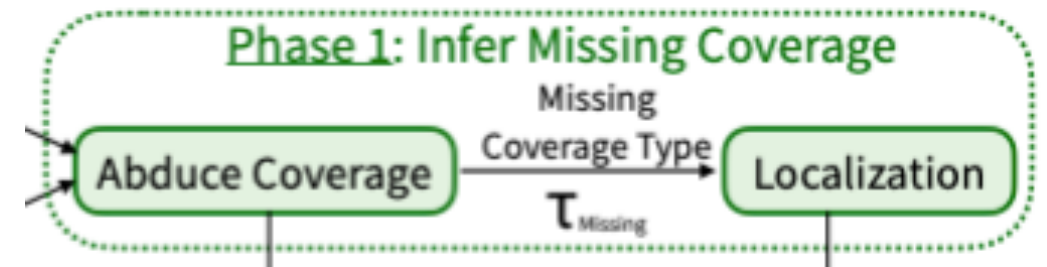
Users provide a **Coverage Specification** and an **incomplete generator**

```
even_list_gen: {s:int | v >= 0}  
-> [int list |  
    all_evens(v) && 0 < len(v) <= s+1]
```

```
let rec even_list_gen  
  (s : int) : int list =  
    if s == 0 then Err  
    else Err
```



Setting up Synthesis



Phase One: Augments the sketch with **typed holes**

```
even_list_gen: {s:int | v >= 0}  
-> [int list |  
    all_evens(v) && 0 < len(v) <= s+1]
```

```
let rec even_list_gen  
  (s : int) : int list =  
    if s == 0 then Err  
    else Err
```

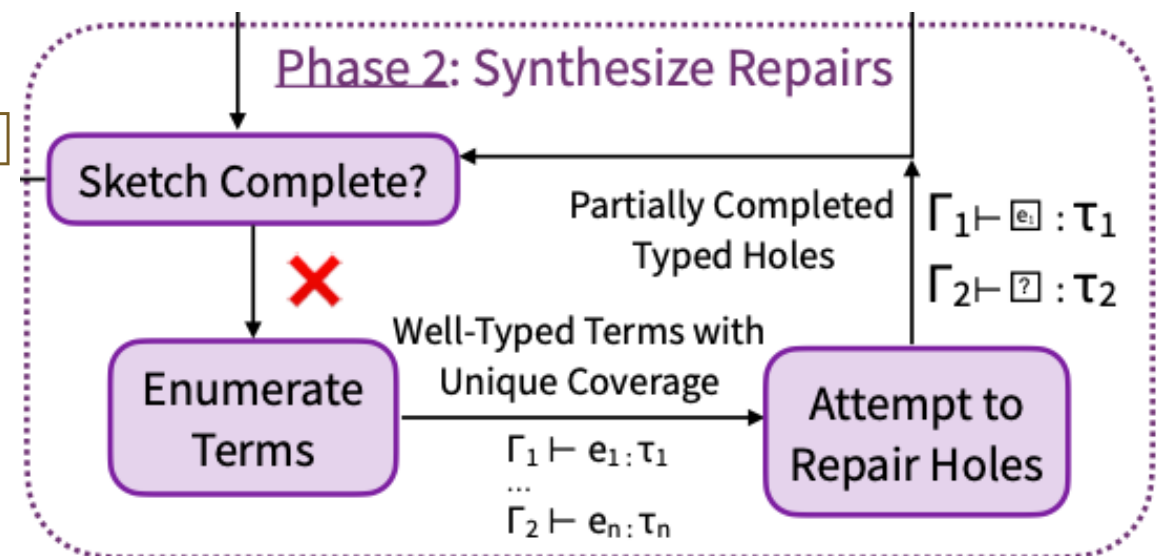
Phase 1

```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then  $\Gamma_1 \vdash \square : \tau_{\text{missing}_1}$   
  else  $\Gamma_2 \vdash \square : \tau_{\text{missing}_2}$ 
```

Setting up Synthesis

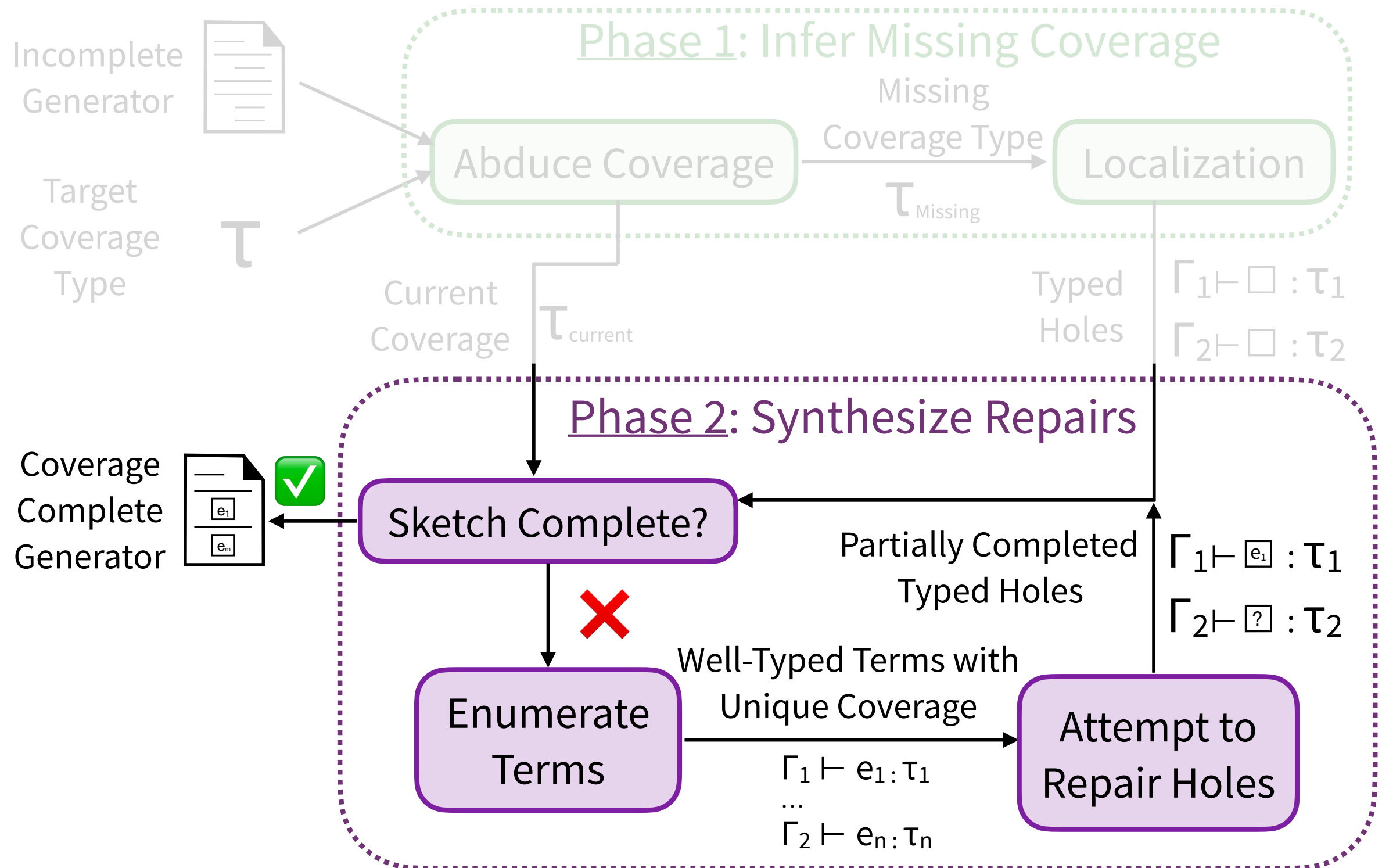
Phase Two: **Component-based Synthesis** to fill holes

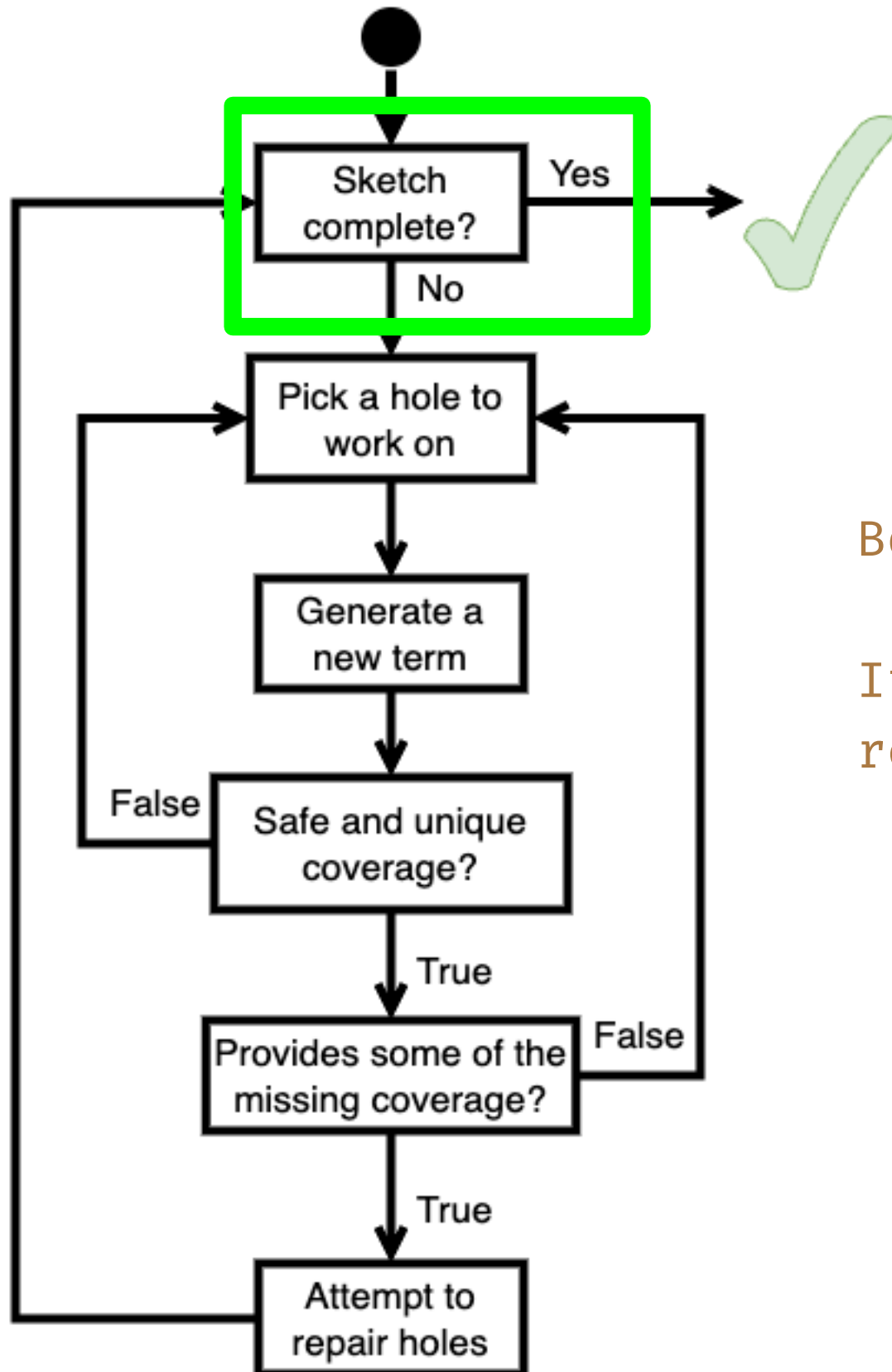
```
even_list_gen: {s:int | v >= 0}  
-> [int list |  
    all_evens(v) && 0 < len(v) <= s+1]
```



```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then  $\Gamma_1 \vdash \square : \tau_{\text{missing}_1}$   
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```

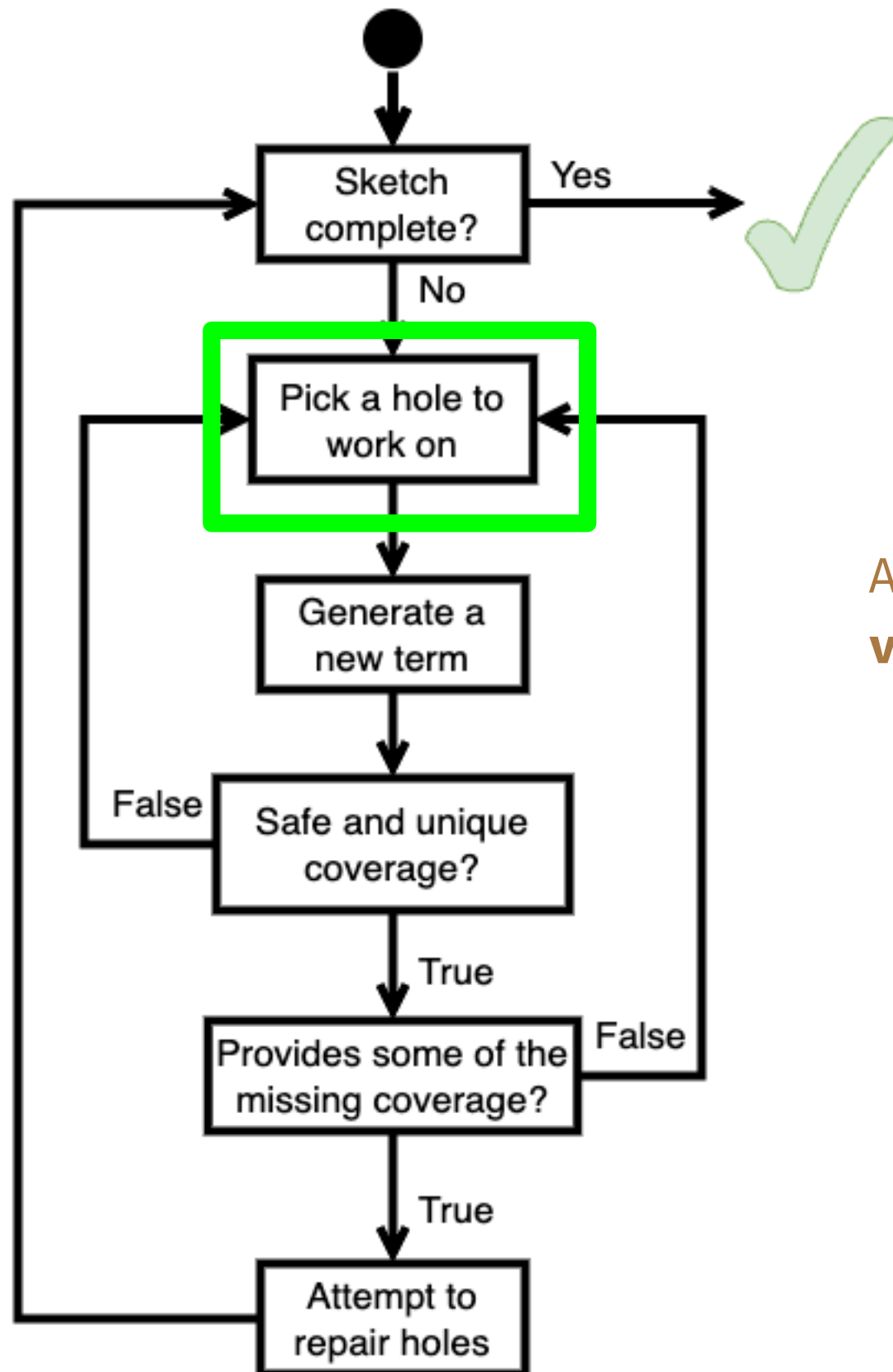
Cobb in a Nutshell





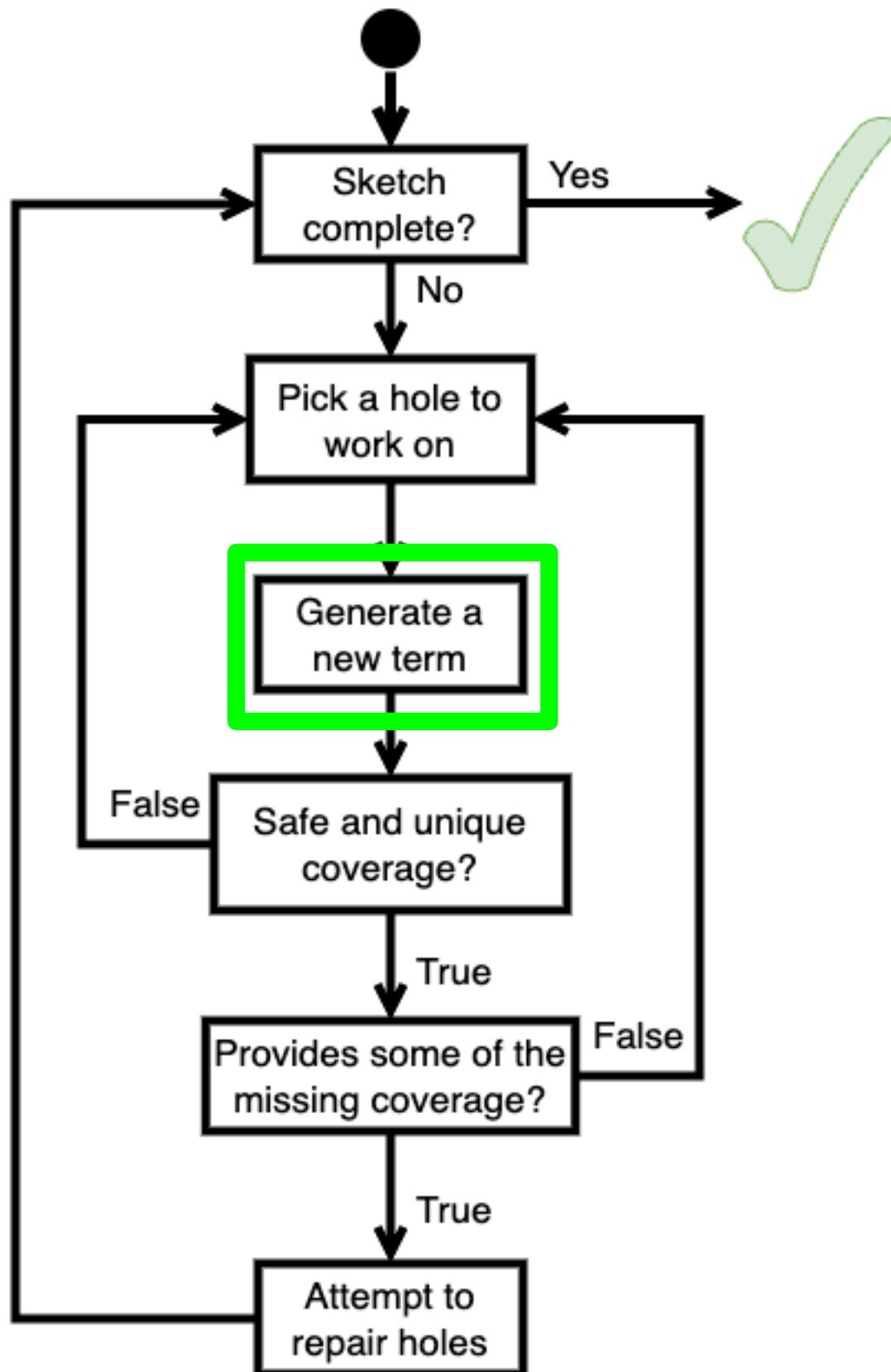
Bottom-up, Cost-based Enumeration

Iterate until sketch is fully repaired or max-bound is hit



A **Typing Context** and set of **local variables** per path

$$\Gamma_1 \vdash \square : \tau_{\text{missing}_1}$$



Term cost: α

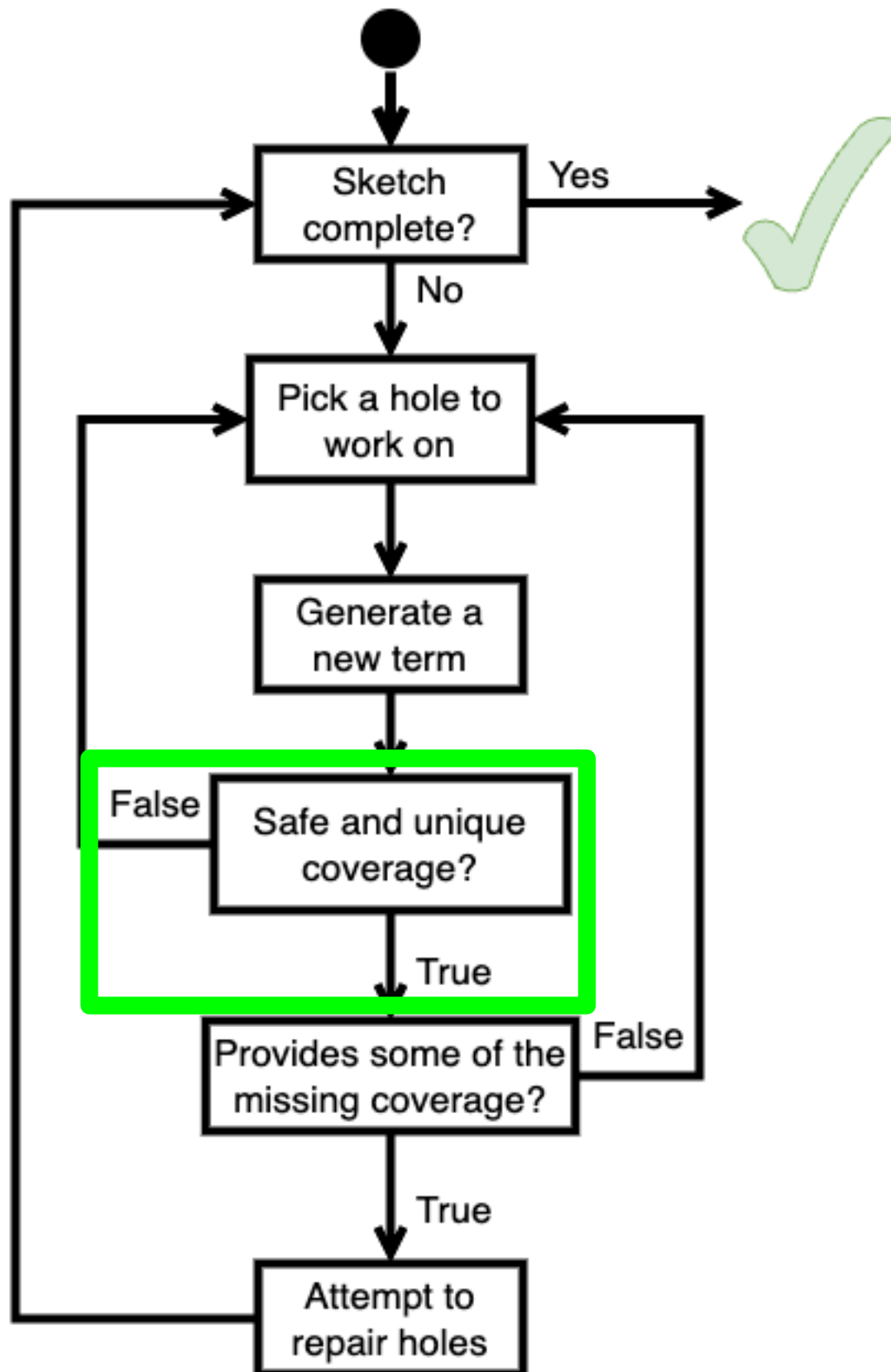
$s-1$

$2 * \text{int_gen}()$

$2 * s$

Small terms that are
initially enumerated

Enumerate Terms



Term cost: α

$\Gamma_1 \vdash s-1 : \tau_1$ ✓

$\Gamma_1 \vdash 2*\text{int_gen}() : \tau_2$ ✓

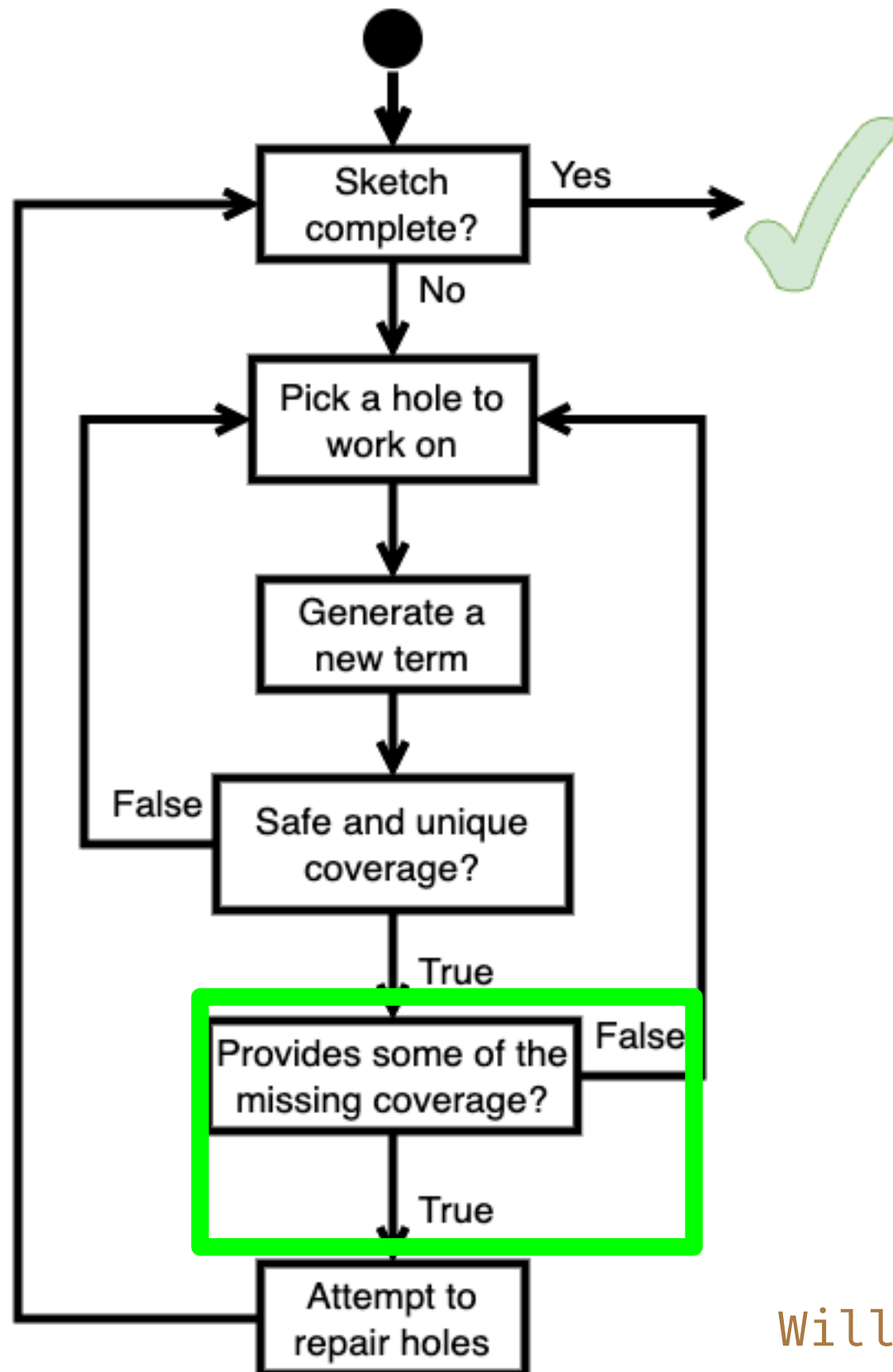
$\Gamma_1 \vdash 2*s : \tau_3$ ✓

Infer a coverage type for each term

Typecheck: safe

Subtyping check: unique

Enumerate Terms



Term cost: α

$\Gamma_1 \vdash s-1 : \tau_1$

$\Gamma_1 \vdash 2*\text{int_gen}() : \tau_2$

$\Gamma_1 \vdash 2*s : \tau_3$

Subtype of missing coverage?

$\tau_{\text{missing_1}} <: \tau_1$



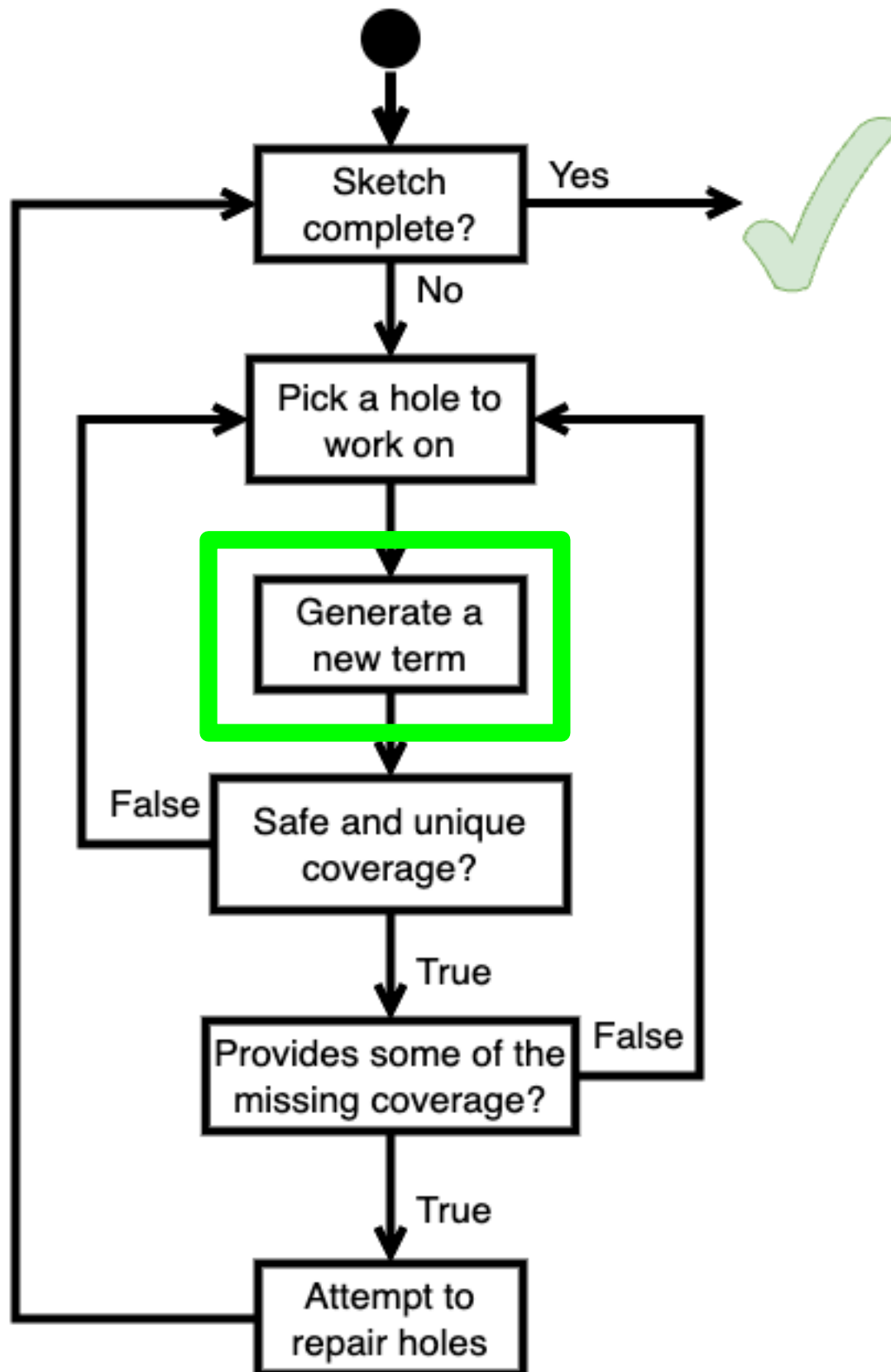
$\tau_{\text{missing_1}} <: \tau_2$



$\tau_{\text{missing_1}} <: \tau_3$



Will be used to enumerate larger terms



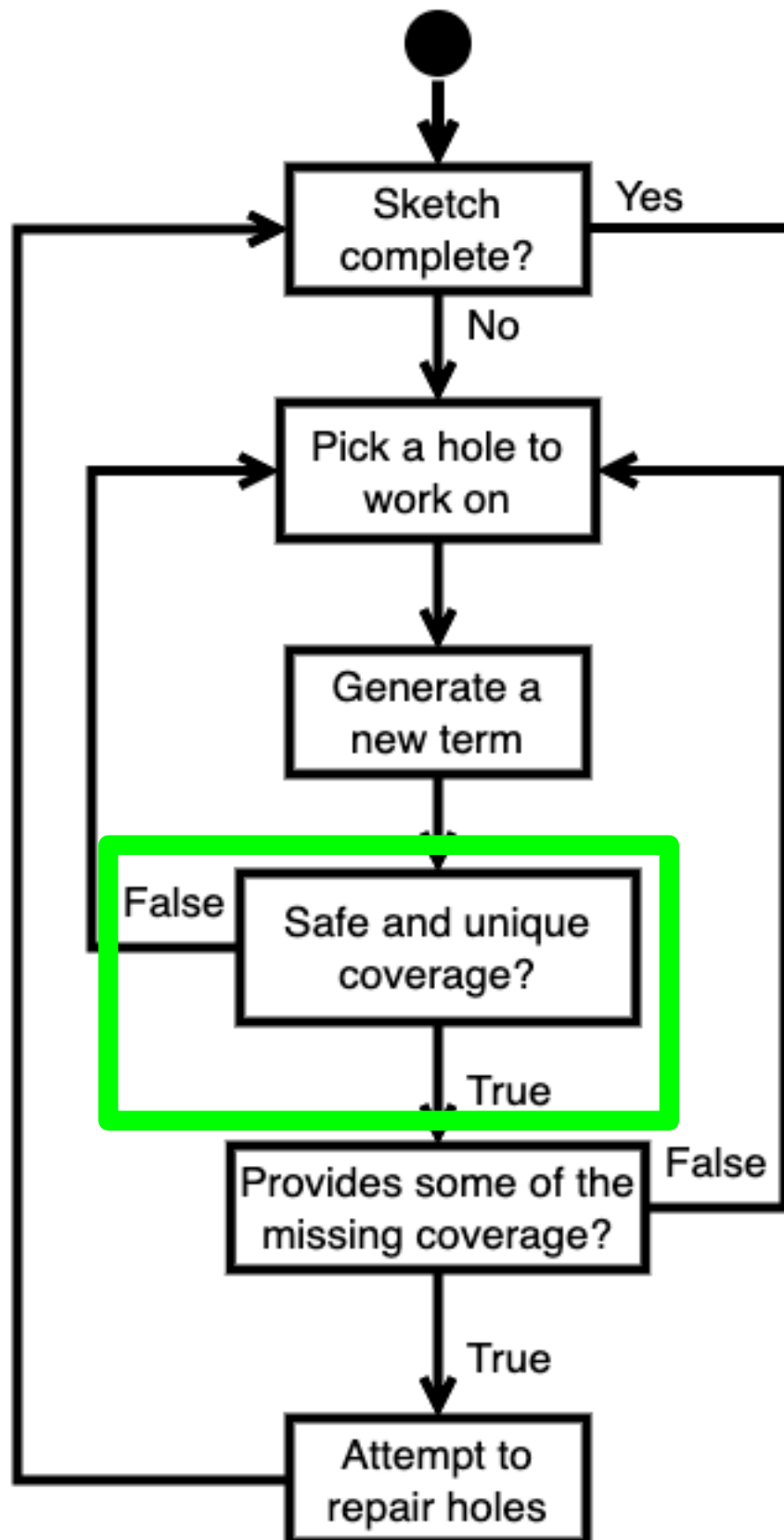
Term cost: α'

`even_list_gen(0)`

`even_list_gen(int_gen())`

`even_list_gen(s - 1)`

Recursive terms are
prioritized



Term cost: α'

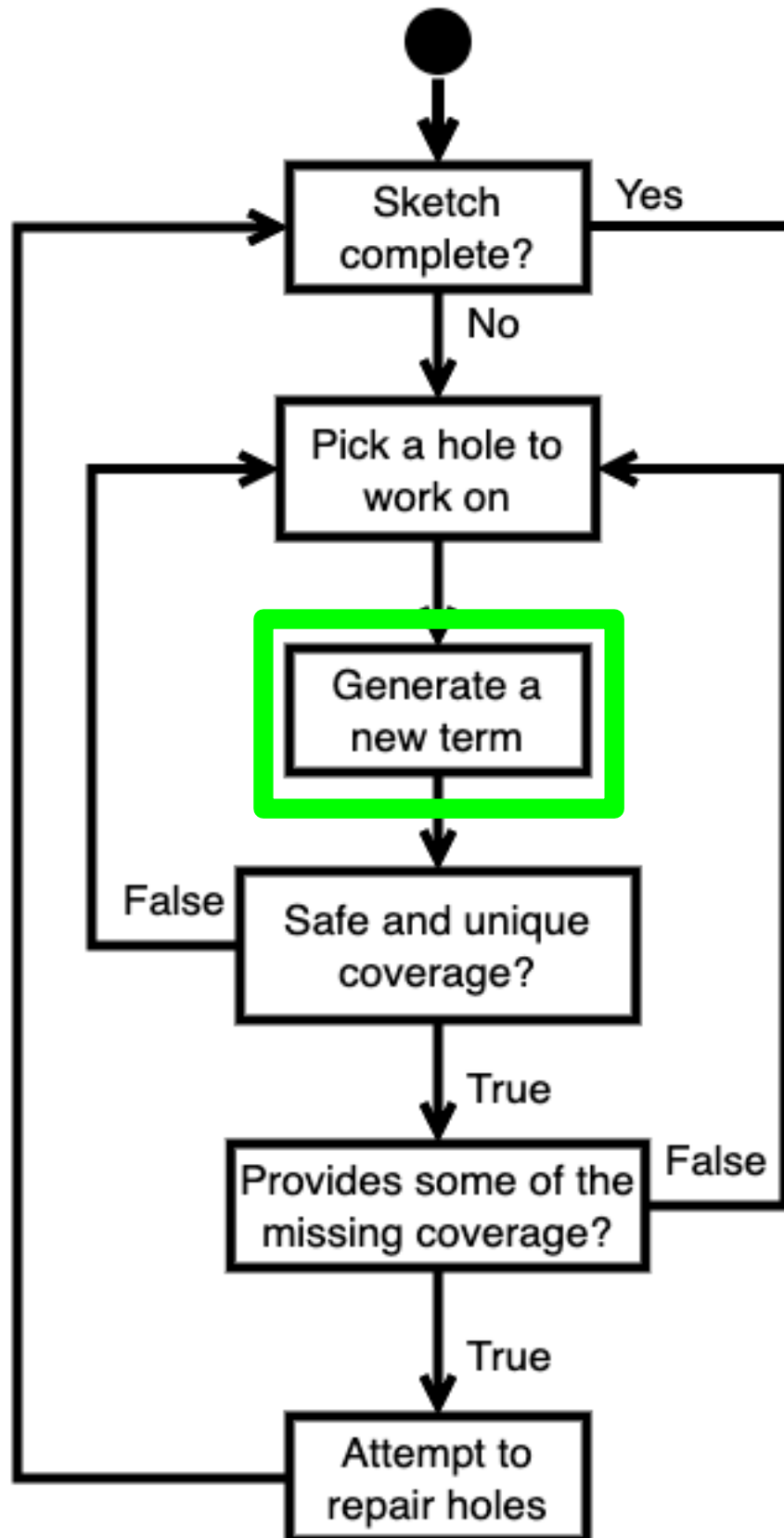
$\Gamma_1 \vdash \text{even_list_gen}(0) : \tau_4$ ❌

$\Gamma_1 \not\vdash \text{even_list_gen}(\text{int_gen}()) : \tau_5$ ❌

$\Gamma_1 \vdash \text{even_list_gen}(s - 1) : \tau_6$ ✅

Some terms are rejected for being **unsafe** or “**coverage equivalent**”

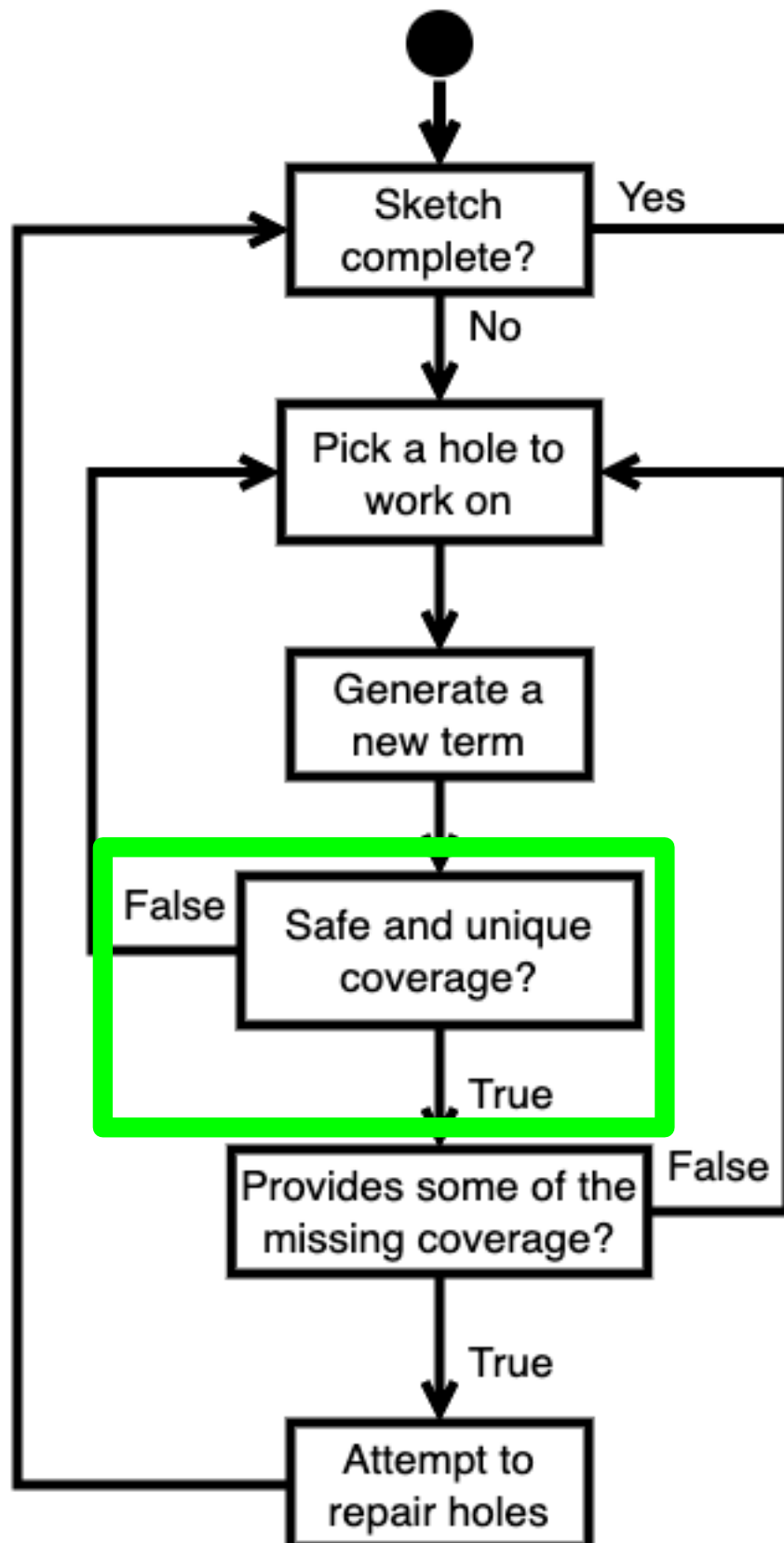
- **int_gen()** not less than s
- **even_list_gen(0)** equivalent to the base case



Term cost: α''

$2 * \text{int_gen}() :: []$

$2 * \text{int_gen}() :: \text{even_list_gen}(s-1)$

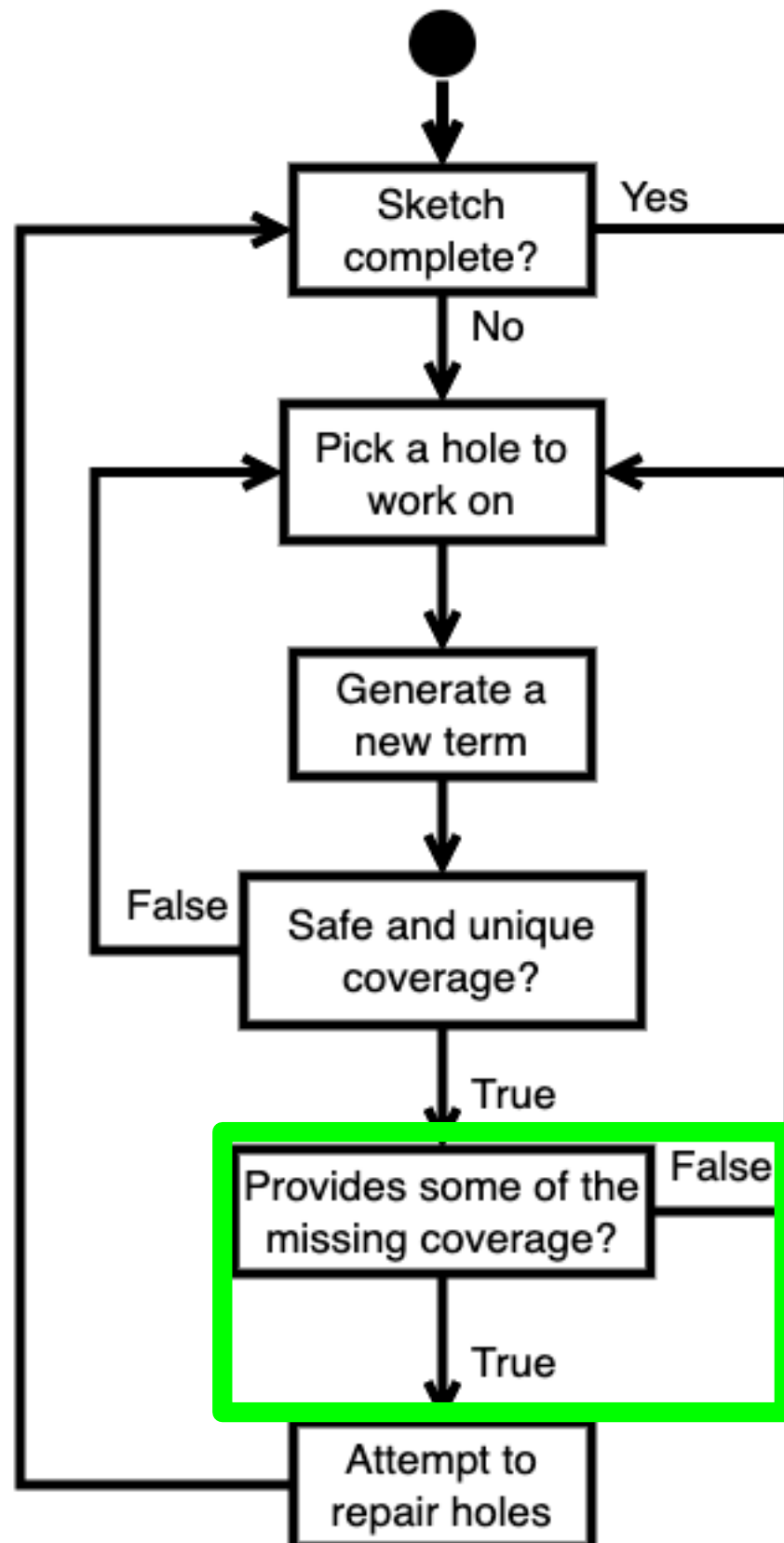


Term cost: α''

$\Gamma_1 \vdash 2 * \text{int_gen}() :: [] : \tau_7$ ✓

$\Gamma_1 \vdash 2 * \text{int_gen}() ::$ ✓

$\text{even_list_gen}(s-1) : \tau_8$



Term cost: α''

$\Gamma_1 \vdash 2 * \text{int_gen}() :: [] : \tau_7$ ✓

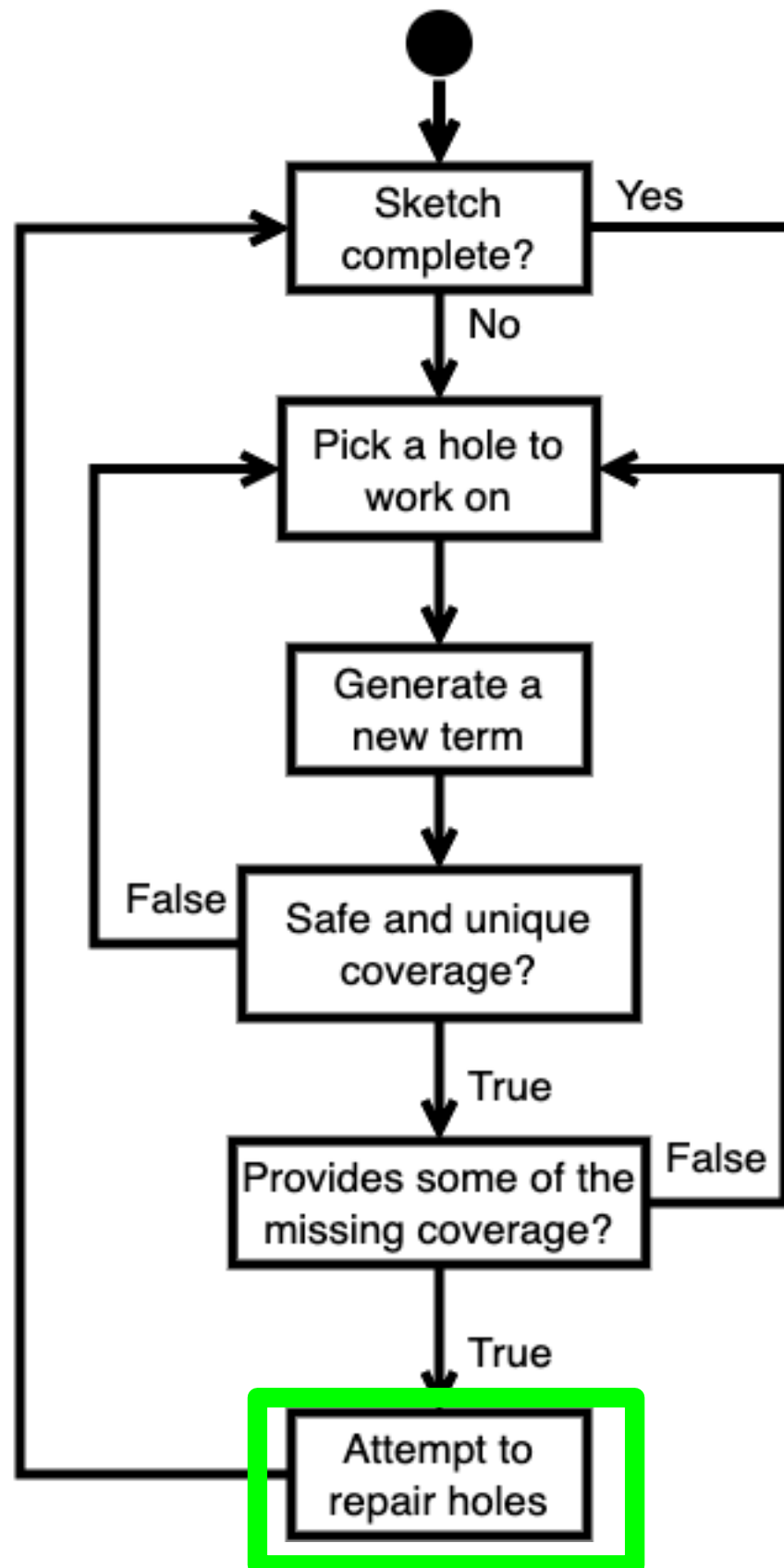
$\Gamma_1 \vdash 2 * \text{int_gen}() ::$ ✓

$\text{even_list_gen}(s-1) : \tau_8$

$\tau_{\text{missing}_1} <: \tau_7$ ✓

$\tau_{\text{missing}_1} <: \tau_8$ ✓

Some terms are a **sub-type** of our **missing coverage**, thus could repair the hole



Term cost: α''

$\Gamma_1 \vdash 2 * \text{int_gen}() :: [] : \tau_7$ ✓

$\Gamma_1 \vdash 2 * \text{int_gen}() ::$ ✓

$\text{even_list_gen}(s-1) : \tau_8$

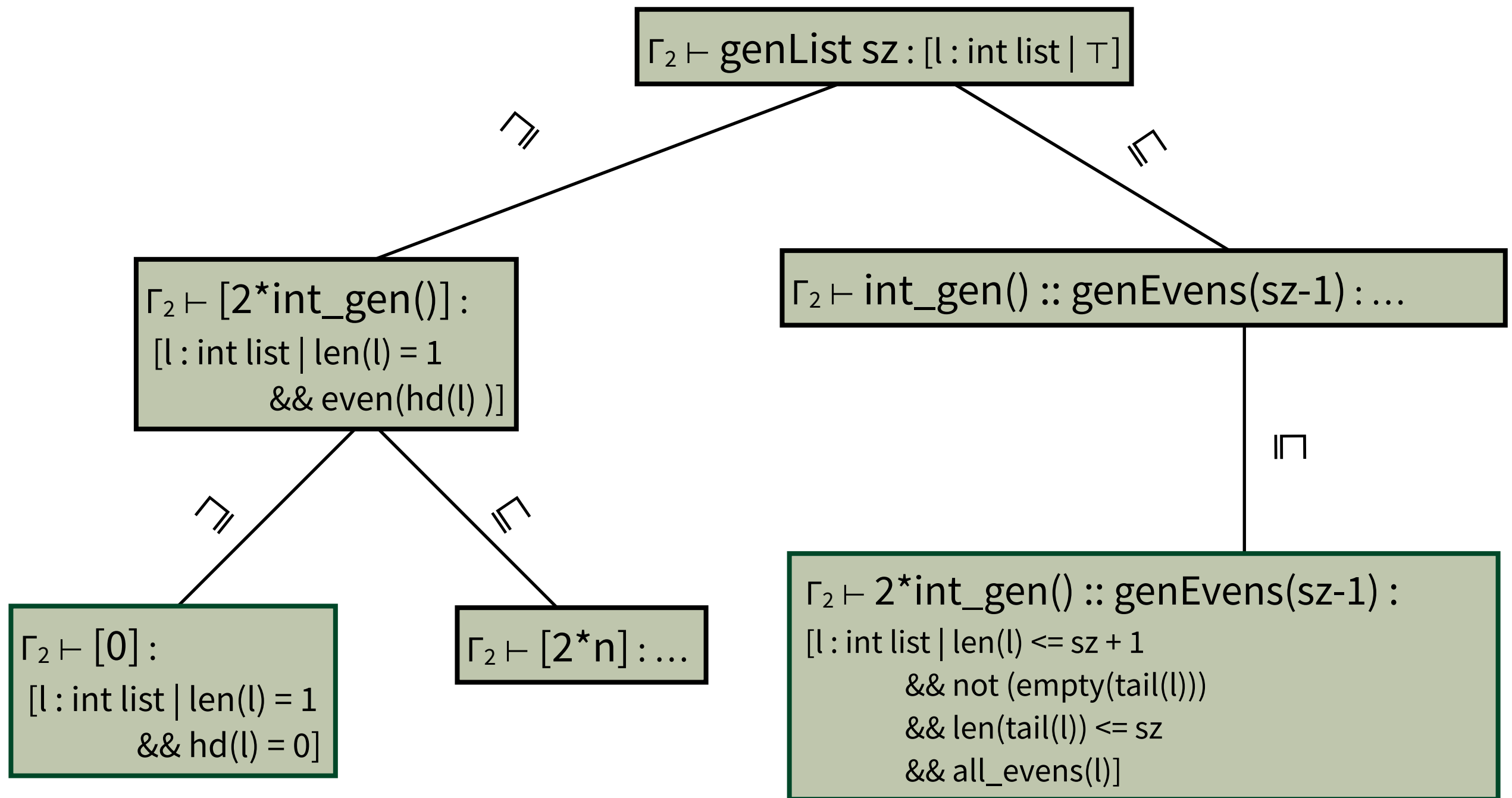
$\tau_{\text{missing}_1} <: \tau_7$ ✓

$\tau_{\text{missing}_1} <: \tau_8$ ✓

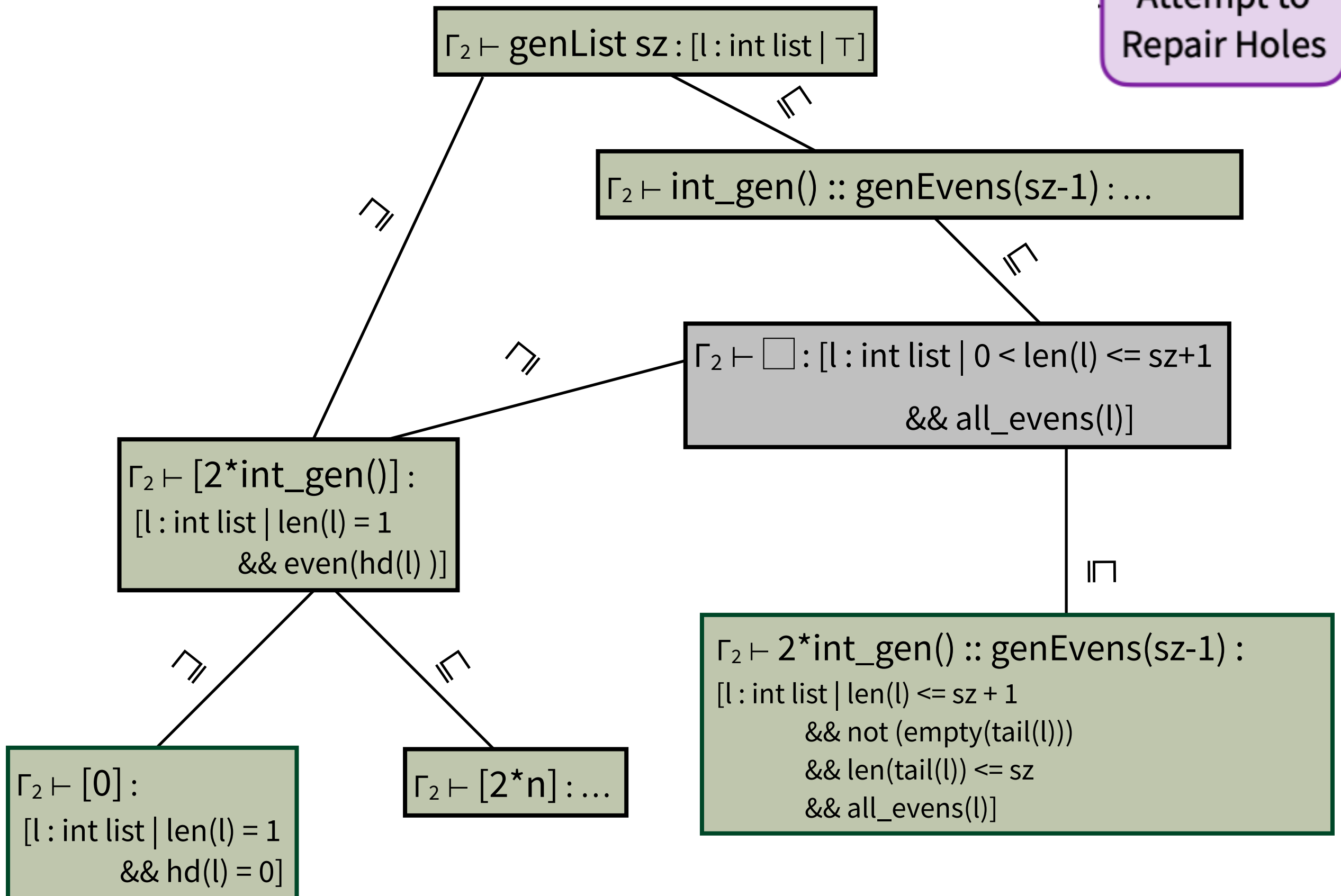
Some terms are a **sub-type** of our **missing coverage**, thus could repair the hole

Extract the optimal repair

Attempt to
Repair Holes



Attempt to
Repair Holes



That was cool and all but show
me some numbers!

Evaluation: Benchmarks

4.5k lines of  **OCaml** (<https://github.com/Pat-Lafon/Cobb>)

Evaluate Cobb on a variety of :

- Datatypes: Lists, Trees, Red-BlackTrees, Simply Typed Lambda Calculus
- Properties: Sized, Sorted, Duplicate, Unique, Valid, Well-Typed
- Typing contexts: Branch conditions, Local variables

Evaluation: Benchmarks

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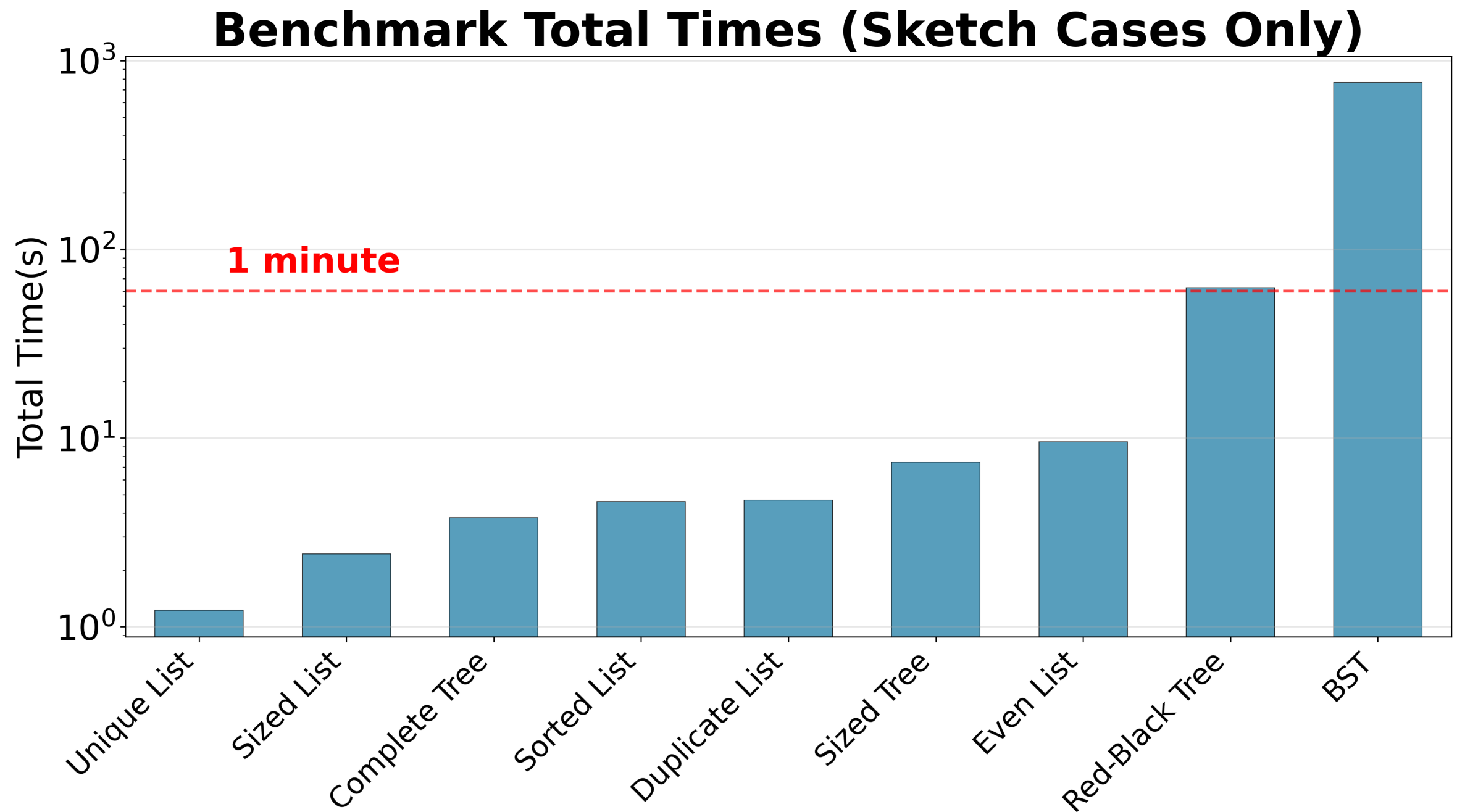
- Datatypes: Lists, Trees, Red-BlackTrees, Simply Typed Lambda Calculus
- Properties: Sized, Sorted, Duplicate, Unique, Valid, Well-Typed
- Typing contexts: Branch conditions, Local variables

Using benchmarks of generators from the literature, created a set of incomplete generator variations

A “Sketch” is then the most general form, only simple control flow and holes

Evaluation: Cobb is pretty fast!

Cobb satisfies the precise coverage < 1 minute



Evaluation: Cobb vs **Coverage-only** repair?

Trivial solution: the default generator is always coverage complete!

```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then  $\Gamma_1 \vdash \square : \tau_{\text{missing}_1}$   
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```

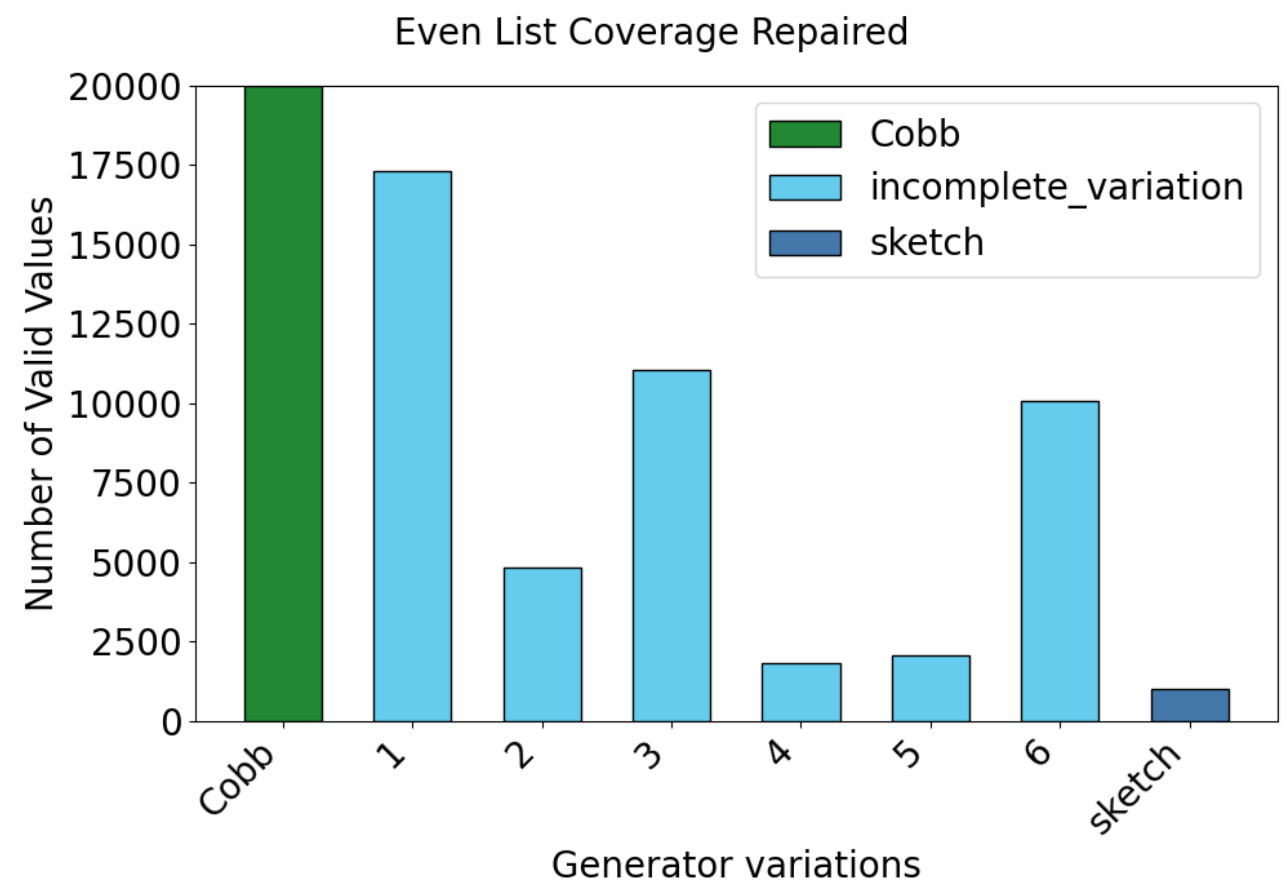
```
let rec even_list_gen (s : int) : int list =  
  if s == 0 then default_list_gen ()  
  else default_list_gen ()
```

Evaluation: Cobb vs **Coverage-only** repair?

Trivial solution: the default generator is always coverage complete!

Count how many outputs are valid (**higher** is better)

```
let rec even_list_gen
  (s : int) : int list =
  if s == 0 then
    default_list_gen ()
  else
    default_list_gen ()
```



See Paper(or come talk to me) for more!

- Simply Typed Lambda Calculus evaluation
- More detailed evaluation numbers/analysis
- Comparison with a safety-based repair strategy
- Comparison with dynamic test input generation
 - Dynamic Constraint solving is at least an order of magnitude slower
- Sensitivity to the set of user-provided components
- Soundness Theorem

Conclusion

- Problem: Users don't want to write input generators

	Safety	Coverage
Type-Based Verification	Liquid Haskell Flux RefinedRust RefinedC refined(scala) ...	Poirot
Type-Based Synthesis	Synquid Cobalt ReSyn SuSLik ...	Cobb (This work)



Conclusion

- Problem: Users don't want to write input generators
- Solution: Coverage Type-Guided Program Repair
 - Phase 1: Characterizes missing coverage
 - Phase 2: Uses coverage to guide repair

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Conclusion

- Problem: Users don't want to write input generators
- Solution: Coverage Type-Guided Program Repair
 - Phase 1: Characterizes missing coverage
 - Phase 2: Uses coverage to guide repair
- Results: Efficient, one-time synthesis task
- <https://github.com/Pat-Lafon/Cobb>

	Safety	Coverage
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