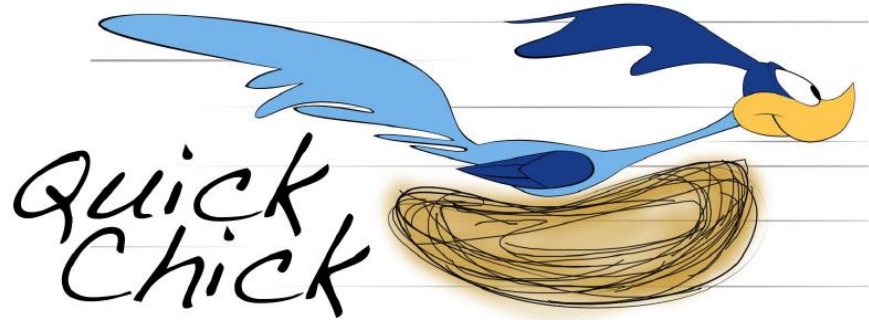


QuickChick : Property-Based Testing in Coq



POPL 2019

Tutorial Fest

14/01/2019

Leonidas Lampropoulos

Benjamin C. Pierce



Software Foundations Volume

SOFTWARE FOUNDATIONS
VOLUME 4

QuickChick: Property-Based Testing in Coq

Leonidas Lampropoulos
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High-level view of workflow

Theorem preservation :=

forall $e \tau \Gamma$,

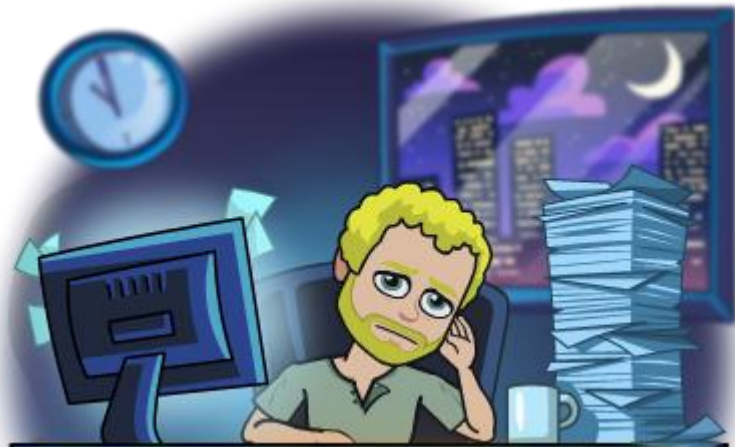
$\Gamma \vdash e : \tau \rightarrow e \Rightarrow e' \rightarrow \Gamma \vdash e' : \tau.$

High-level view of workflow

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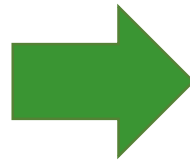
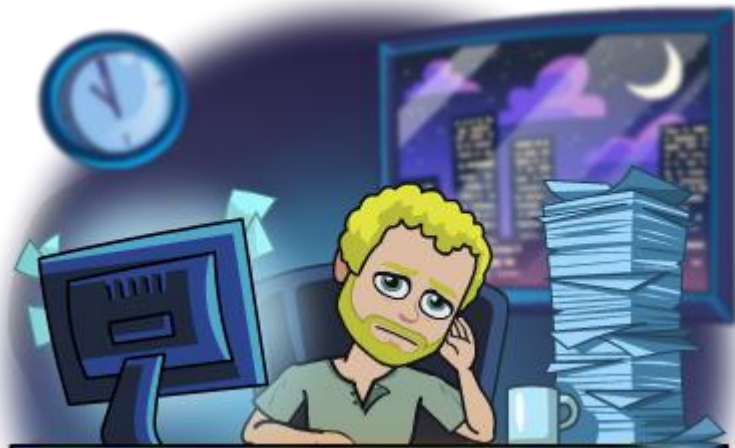


High-level view of workflow

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A better workflow

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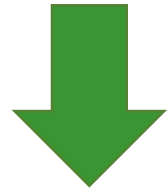
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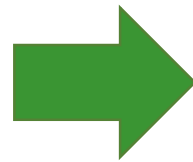
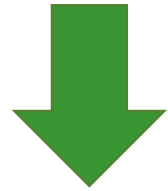


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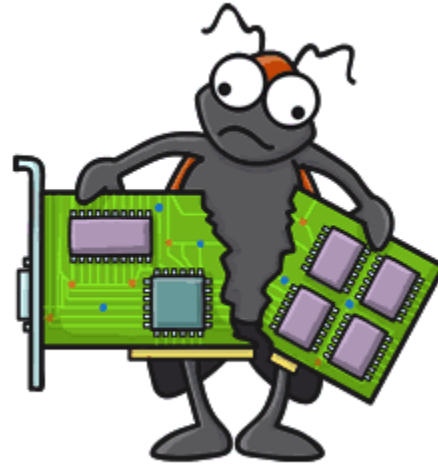



A better workflow

Theorem preservation :=
forall $e \tau \Gamma$,
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Bugs are everywhere



Testing  Verification

Testing

- Excellent at discovering bugs
 - [CSmith] : More than 400 bugs in C compilers (GCC, LLVM)
 - [Palka et al. '11] : Bugs in GHC's strictness analyzer
- Cannot guarantee their absence
 - “Testing shows the presence, not the absence of bugs” - Dijkstra

Verification

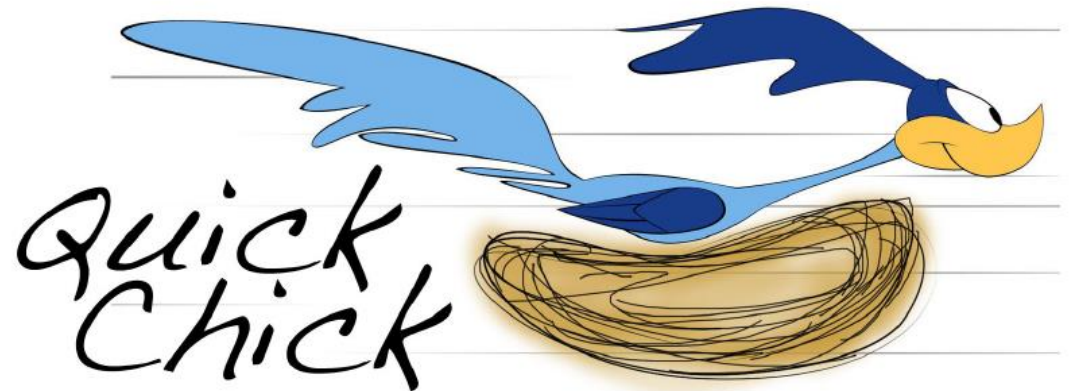
- Strong formal guarantees
- Recent success stories
 - [CompCert] : Verified optimizing C compiler
 - [CertiKOS] : Extensible architecture for certified OS kernels
- ...but still very expensive

Testing and Verification

- Already present in many proof assistants
 - Isabelle [Berghofer 2004, Bulwahn 2012]
 - Agda [Dybjer et al 2003]
 - ACL2 [Chamarthi et al 2011]

Testing in Coq

- Already present in many proof assistants
 - Isabelle [Berghofer 2004, Bulwahn 2012]
 - Agda [Dybjer et al 2003]
 - ACL2 [Chamathi et al 2011]
- QuickChick
 - Coq port of Haskell QuickCheck
 - On steroids! [*ITP 2015, POPL 2018*]



Roadmap

- Property-based testing overview
- Case Study: Expression Compiler
 - Typeclasses and automation
 - Properties
 - Batch execution
 - Mutation Testing
- Beyond Automation - Generators
- Open Research!

Overview of property-based testing

```
Theorem preservation :=  
  forall e T Γ,  
    Γ |- e : T -> e => e' -> Γ |- e' : T.
```

```
QuickChick preservation.
```


Overview of property-based testing

Decidable
Property

Theorem preservation :=
forall $e \tau \Gamma$,
 $\Gamma \vdash e : \tau \rightarrow e \Rightarrow e' \rightarrow \Gamma \vdash e' : \tau$.

QuickChick preservation.

Overview of property-based testing

Generate random
inputs

Decidable
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Theorem preservation :=
forall e T Γ ,
 $\Gamma \vdash e : T \rightarrow e \Rightarrow e' \rightarrow \Gamma \vdash e' : T$.

QuickChick preservation.

Overview of property-based testing

Generate random inputs

Decidable Property

```
Theorem preservation :=  
  forall e T Γ,  
    Γ |- e : T -> e => e' -> Γ |- e' : T.
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```
QuickChick preservation.
```

Test result

Overview of property-based testing

Generate random inputs

Decidable Property

```
Theorem preservation :=  
  forall e T Γ,  
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QuickChick preservation.
```

Test result

success

Overview of property-based testing

Generate random inputs

Decidable Property

```
Theorem preservation :=  
forall e T Γ,  
  Γ |- e : T -> e => e' -> Γ |- e' : T.
```

```
QuickChick preservation.
```

success

Test result

fail

Report counterexample

Demo time!

Arithmetic Expressions

42

$17 + 25$

$(22 - 1) * 2$

Arithmetic Expressions

42

17 + 25

(22 - 1) * 2

```
Inductive exp : Type :=  
  | ANum : nat → exp  
  | APlus : exp → exp → exp  
  | AMinus : exp → exp → exp  
  | AMult : exp → exp → exp.
```


Arithmetic Expressions

42

`ANum 42`

$17 + 25$

`APlus (ANum 17) (ANum 25)`

$(22 - 1) * 2$

`AMult (AMinus (ANum 22) (ANum 1)) (ANum 2)`

```
Inductive exp : Type :=
| ANum : nat → exp
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Arithmetic Expressions - Printing

```
Inductive exp : Type :=  
  | ANum : nat → exp  
  | APlus : exp → exp → exp  
  | AMinus : exp → exp → exp  
  | AMult : exp → exp → exp.
```

Arithmetic Expressions - Printing

```
Fixpoint show_exp e : string :=
  match e with
  | ANum e => "ANum " ++ show_nat e
  | APlus e1 e2 =>
    "APlus (" ++ show_exp e1 ++ ") (" ++ show_exp e2 ++ ")"
  | ...
```

```
Inductive exp : Type :=
  | ANum : nat → exp
  | APlus : exp → exp → exp
  | AMinus : exp → exp → exp
  | AMult : exp → exp → exp.
```

Arithmetic Expressions - Printing

Explicitly
remember how to
print nats

```
Fixpoint show_exp e : string :=  
  match e with  
  | ANum e => "ANum " ++ show_nat e  
  | APlus e1 e2 =>  
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  | AMinus : exp → exp → exp  
  | AMult : exp → exp → exp.
```

The Show TypeClass

```
class Show (A : Type) := { show : A -> string }.
```

The Show TypeClass

Class method

```
class Show (A : Type) := { show : A -> string }.
```

The Show TypeClass

Class method

```
Class Show (A : Type) := { show : A -> string }.
```

```
Instance show_exp : Show exp :=
```

```
{ | show e :=
```

```
  match e with
```

```
  | ANum e => "ANum " ++ show_nat e
```

```
  | APlus e1 e2 =>
```

```
    "APlus (" ++ show e1 ++ ") (" ++ show e2 ++ ")"
```

```
  | ...
```

```
  | }.
```

The Show TypeClass

Class method

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Class Show (A : Type) := { show : A -> string }.
```

```
Instance show_exp : Show exp :=
```

```
{ | show e :=
```

```
  match e with
```

```
  | ANum e => "ANum " ++ show e
```

```
  | APlus e1 e2 =>
```

```
    "APlus (" ++ show e1 ++ " ) (" ++ show e2 ++ " )"
```

```
  | ...
```

```
  | }.
```

Class method

The Show TypeClass

```
class Show (A : Type) := { show : A -> string }.
```

Derive Show for exp.

Arithmetic Expressions - Generation

```
Class Gen (A : Type) := { arbitrary : G A }.
```

Derive Arbitrary for exp.

Arithmetic Expressions - Generalization

Class method

```
Class Gen (A : Type) := { arbitrary : G A }.
```

Derive Arbitrary for exp.

Arithmetic Expressions - Generation

Class method

```
Class Gen (A : Type) := { arbitrary : G A }.
```

Generation Monad

Derive Arbitrary for exp.

Demo Time! - Sampling

Evaluation

```
Fixpoint eval (e : exp) : nat :=
  match e with
  | ANum n => n
  | APlus e1 e2 => (eval e1) + (eval e2)
  | AMinus e1 e2 => (eval e1) - (eval e2)
  | AMult e1 e2 => (eval e1) * (eval e2)
  end.
```

An expression optimizer

Optimizations:

$$0 + x \rightarrow x$$

$$X + 0 \rightarrow x$$

$$X * 0 \rightarrow 0$$

$$X * 1 \rightarrow x$$

$$X - 0 \rightarrow x$$

...

An expression optimizer - Correctness

Optimizations:

$$0 + x \rightarrow x$$

$$x + 0 \rightarrow x$$

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$$x - 0 \rightarrow x$$

...

$e : \text{exp}$

An expression optimizer - Correctness

Optimizations:

$$0 + x \rightarrow x$$

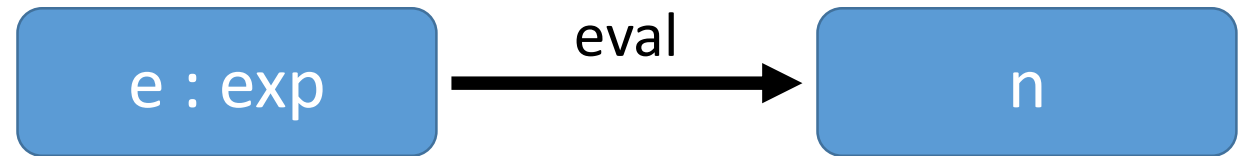
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An expression optimizer - Correctness

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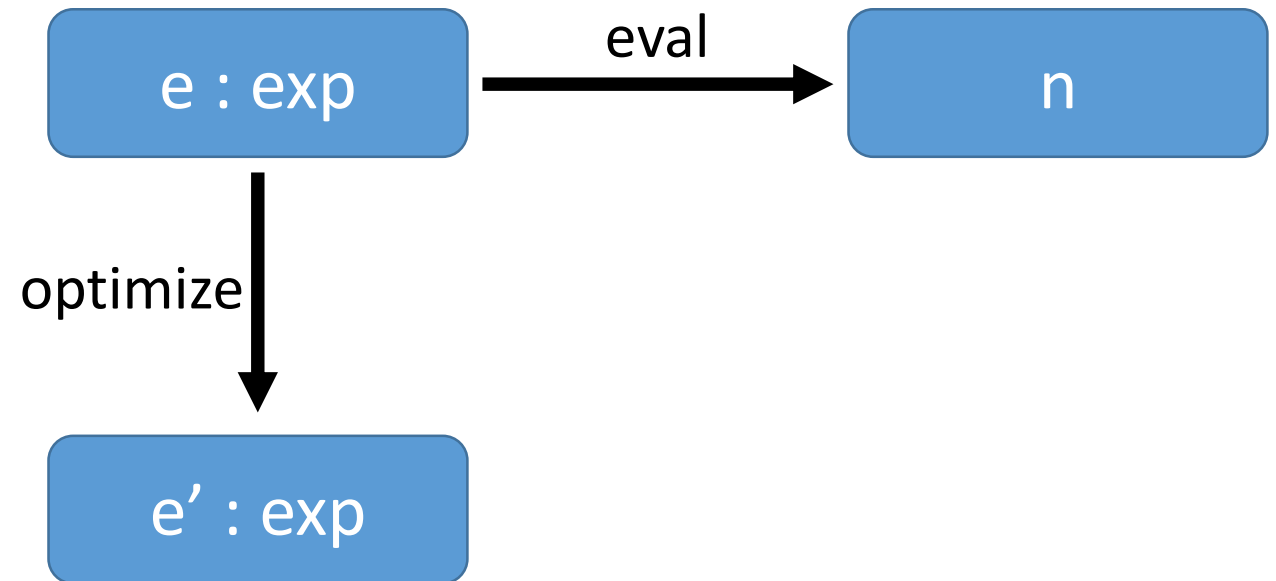
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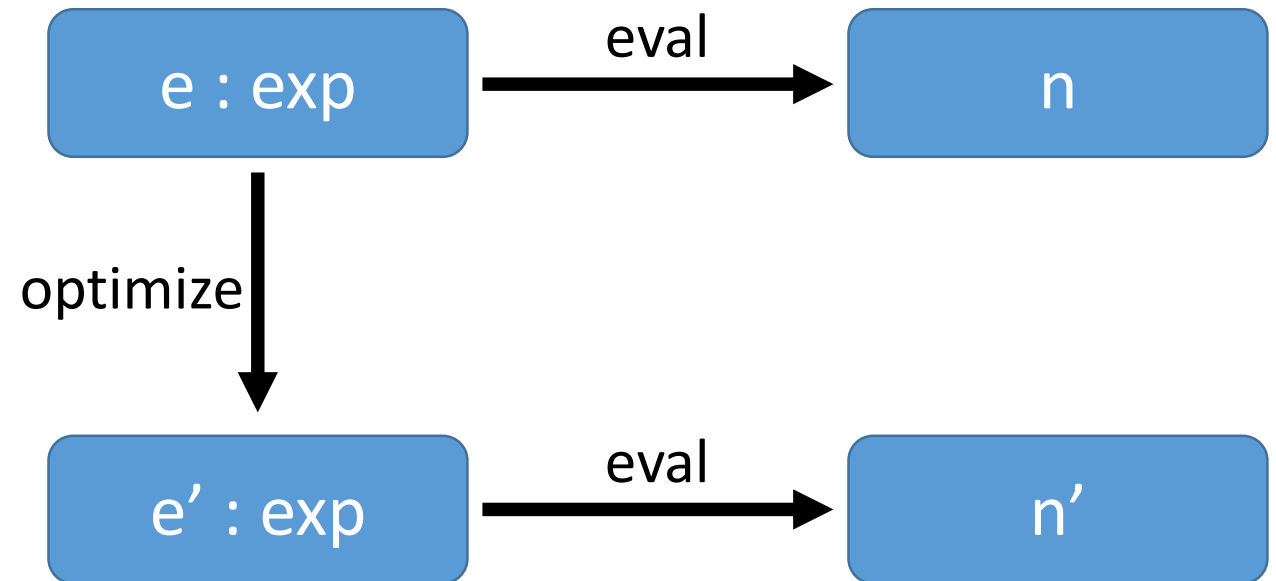
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An expression optimizer - Correctness

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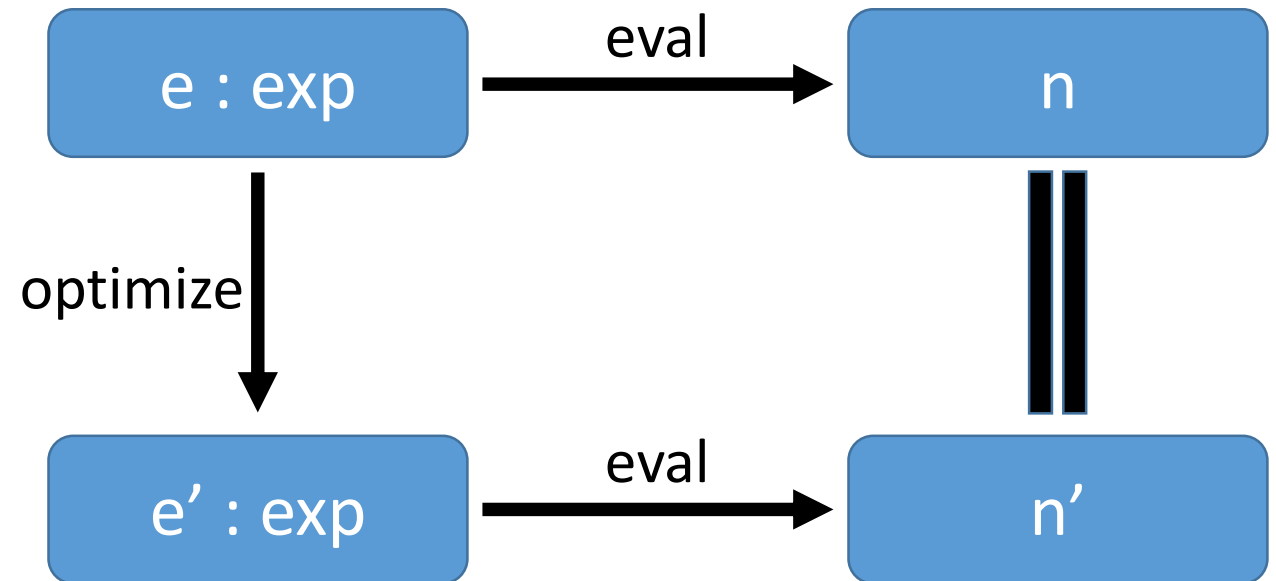
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$$x * 1 \rightarrow x$$

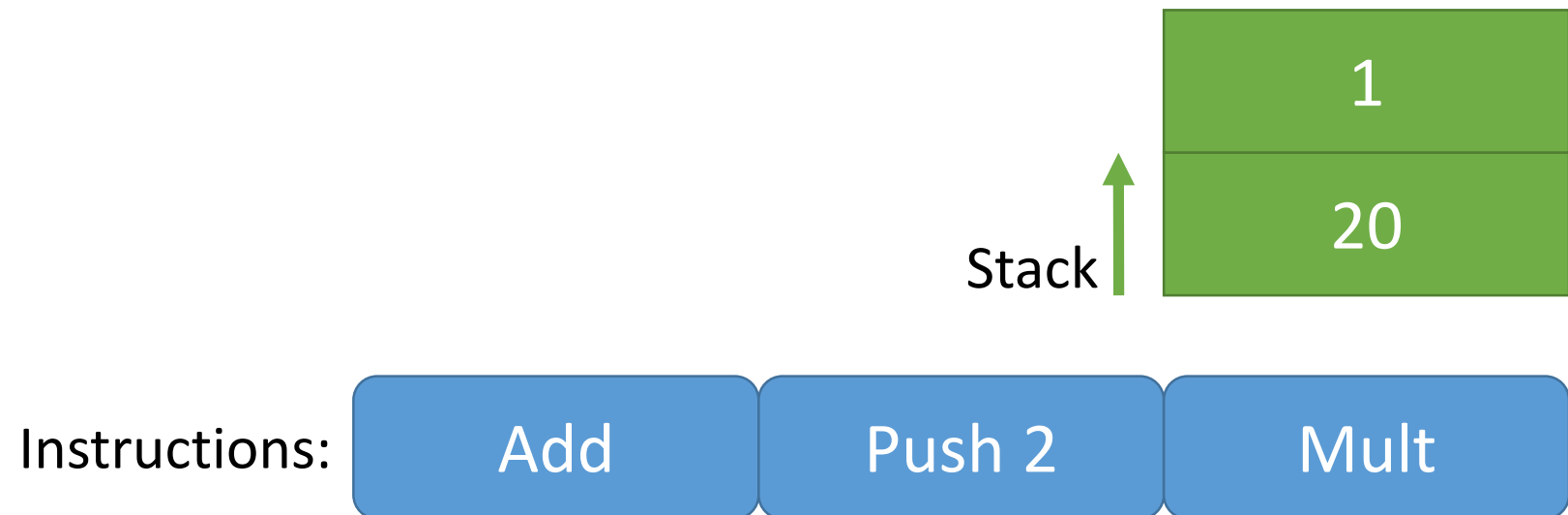
$$x - 0 \rightarrow x$$

...

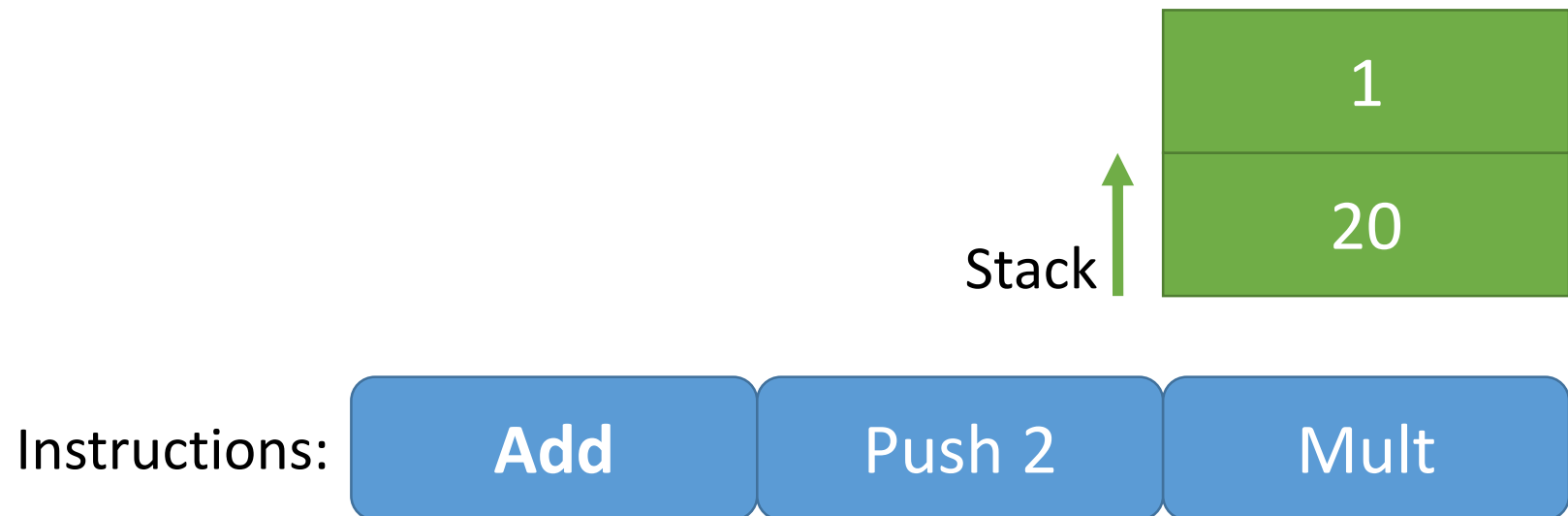


The Stack Machine

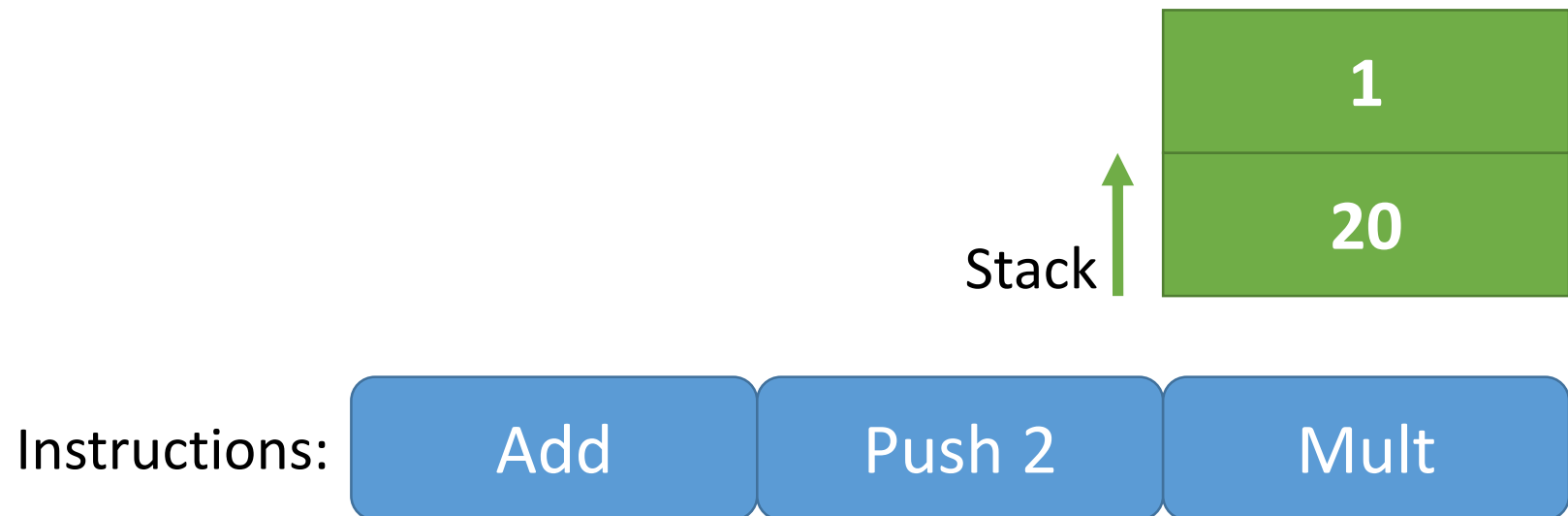
The Stack Machine



The Stack Machine



The Stack Machine



The Stack Machine

Instructions:

Push 2

Mult



The Stack Machine

Instructions:

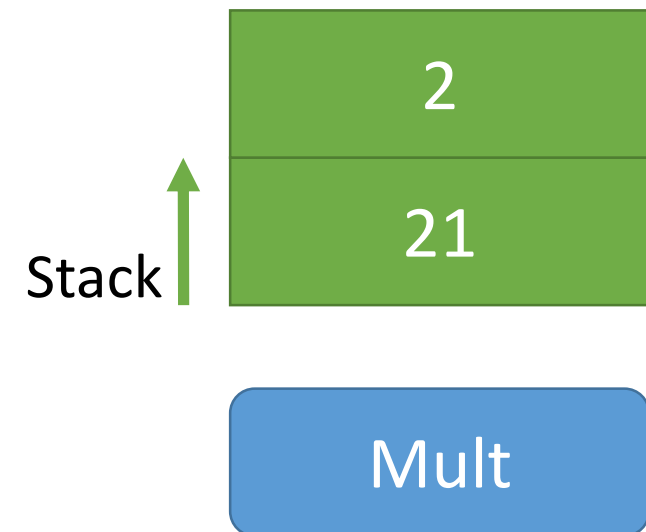
Push 2

Mult



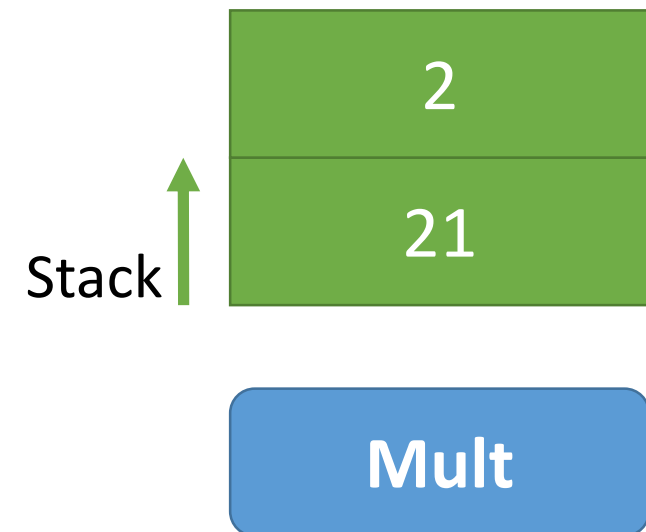
The Stack Machine

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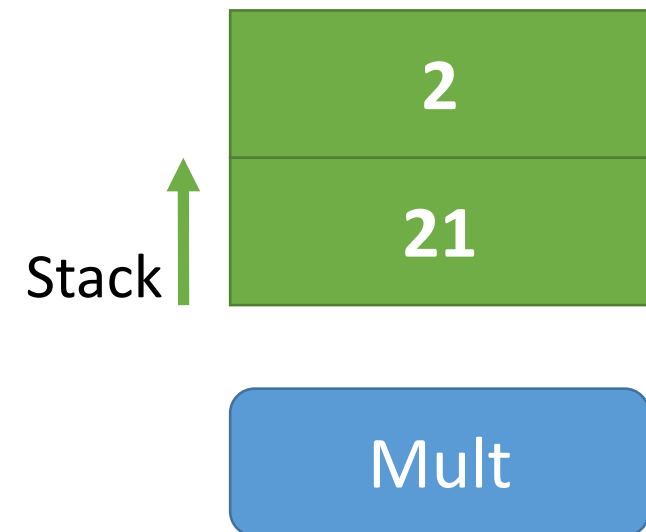
The Stack Machine

Instructions:



The Stack Machine

Instructions:



The Stack Machine



Instructions:

The Compiler

$(20 + 1) * 2$

The Compiler

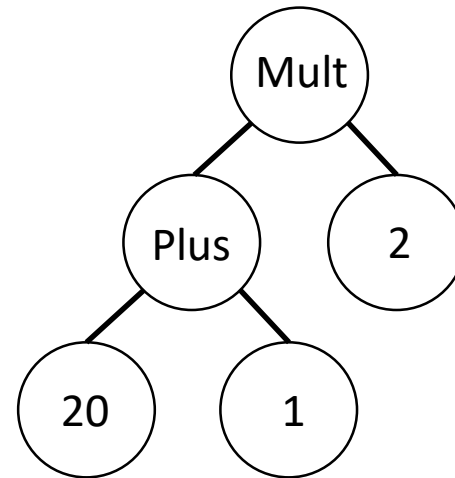
$(20 + 1) * 2$

```
AMuIt (APlus (ANum 20) (ANum 1)) (ANum 2)
```


The Compiler

$(20 + 1) * 2$

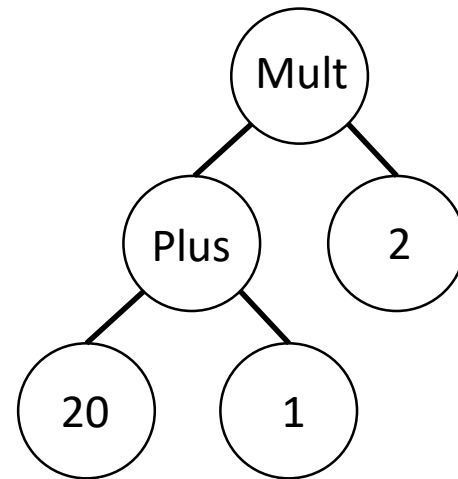
`AMult (APlus (ANum 20) (ANum 1)) (ANum 2)`



The Compiler

$(20 + 1) * 2$

AMult (APlus (ANum 20) (ANum 1)) (ANum 2)



Push 20

Push 1

Add

Push 2

Mult

The Compiler - Correctness

```
Theorem compile_correct :=  
  forall e,  
    [eval e] = execute (compile e []).
```

```
QuickChick compile_correct.
```

The Compiler - Correctness

Dec Typeclass

```
Theorem compile_correct :=  
  forall e,  
    [eval e] = execute (compile e []).
```

```
QuickChick compile_correct.
```

A look under the hood

QuickChick compile_correct.

- Extract “compile_correct” to a single ML file
- Compile (w/ optimizations)
- Run binary

Batch Execution – the command line tool

```
(*! quickChick compile_correct. *)
```

- No overhead when compiling theories
- Use external tool to run all tests with one extraction/compilation!

Mutation testing

How do you know when you're done?

- ... no bugs exist?
- ... testing is not good enough?

Mutation testing

How do you know when you're done?

- ... no bugs exist?
- ... testing is not good enough?

```
Fixpoint compile (e : exp) :=
... (*! *) (*!! MINUS_ASSOC *)
  | AMinus e1 e2 => compile e2 ++ compile e1 ++ [SMinus]
(*!
  | AMinus e1 e2 => compile e1 ++ compile e2 ++ [SMinus]
*) ...
```


The road so far...

$$\forall x.p(x)$$

The road so far...

$$\forall x.p(x)$$

$$\forall x.p(x) \rightarrow q(x)$$

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If x is well
typed

The road so far...

$$\forall x.p(x)$$

$$\forall x.p(x) \rightarrow q(x)$$

If x is well
typed

Then it is either a
value or can take a
step

Properties with preconditions

$$\forall x. p(x) \rightarrow q(x)$$

Properties with preconditions

Generate x

$$\forall x. p(x) \rightarrow q(x)$$

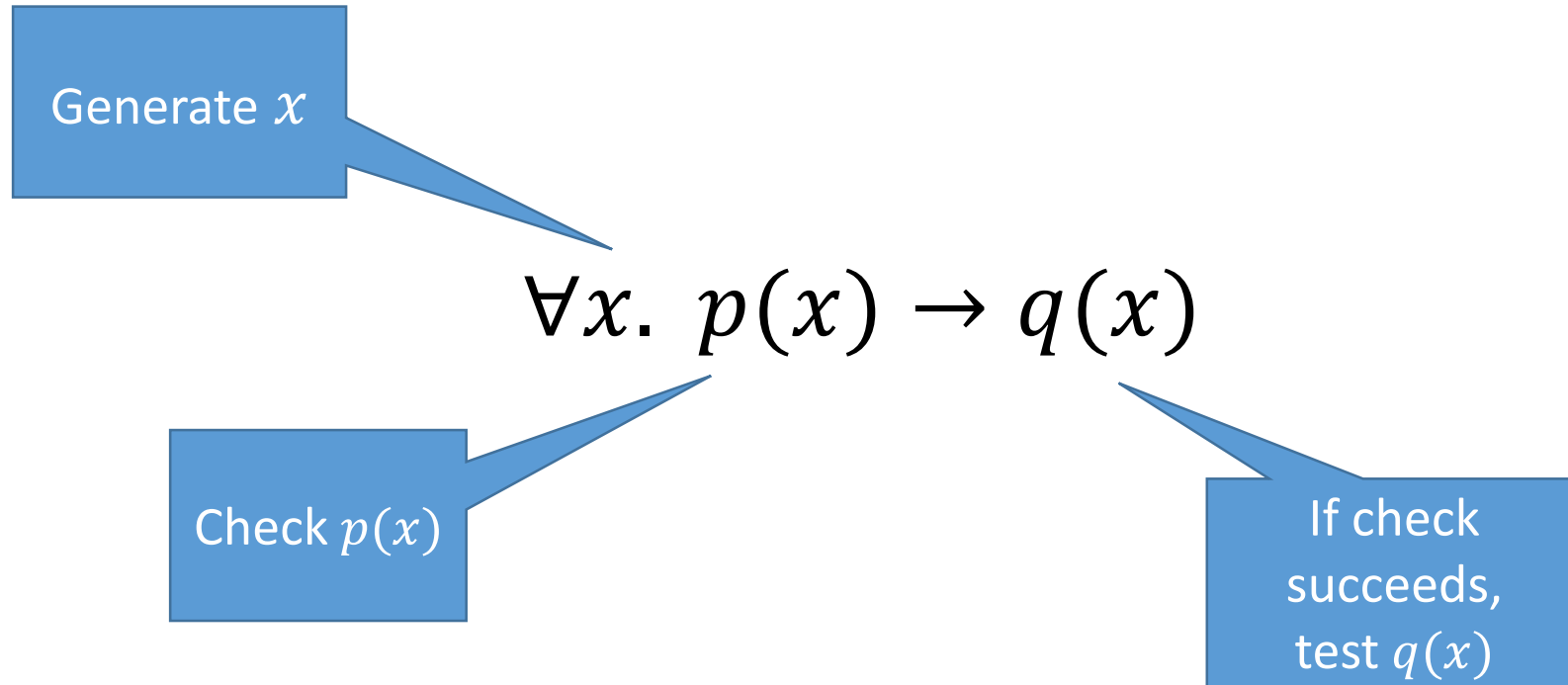
Properties with preconditions

Generate x

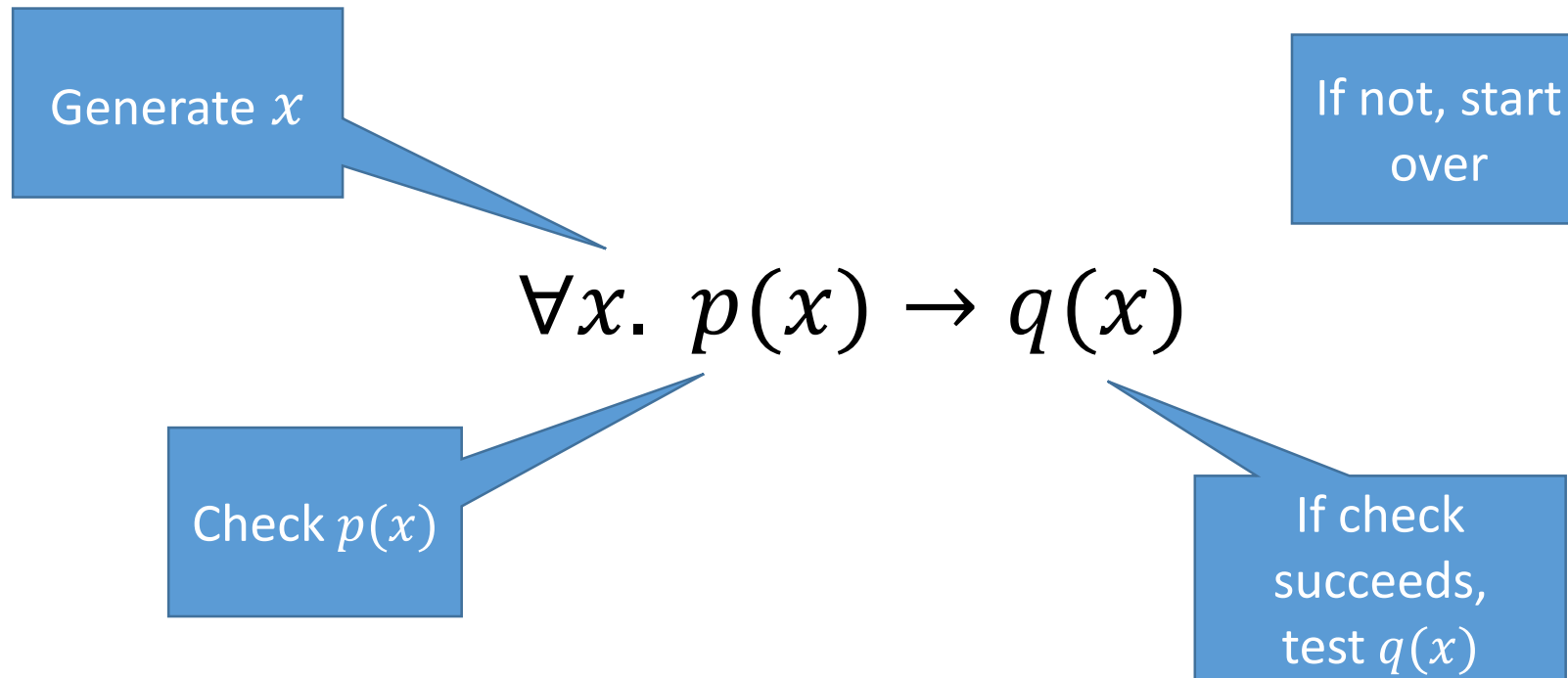
$$\forall x. p(x) \rightarrow q(x)$$

Check $p(x)$

Properties with preconditions



Properties with preconditions



Properties with preconditions

Generate x

If not, start
over

$$\forall x. p(x) \rightarrow q(x)$$

Check $p(x)$

If check
succeeds,
test $q(x)$

SRSLY?!

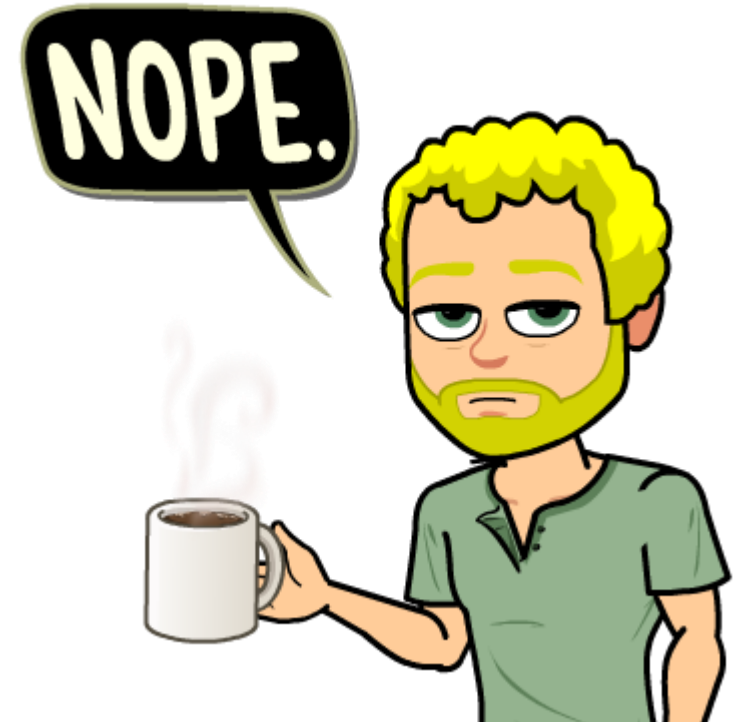


Let's generate well-typed terms!

GOAL: Generate e , such that $(\text{well_typed } e \ T)$ holds for a given T

Take 1 – Generate and test

- Assume we can *decide* whether a term has a given type
- Generate random lambda terms
- Filter out the ill-typed ones



Take 2 – Custom generators

Solution: Write a generator that produces well-typed terms!

The road ahead...

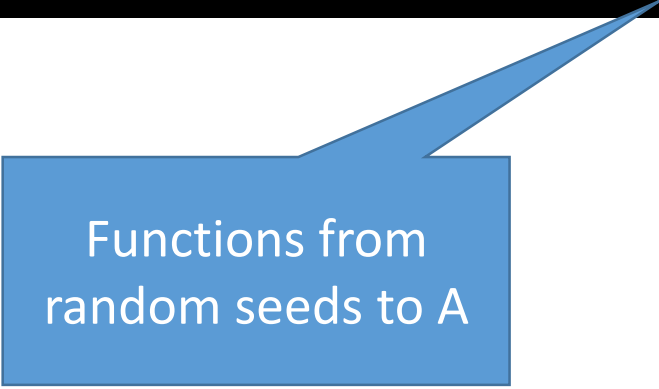
- Generators
 - Generation Monad
 - Primitive Generators
 - Generator Combinators – Trees!
- Properties
 - Decidability
 - The Checkable Class
 - An example: Mirrored Trees
- Open Research!

Generation Overview

```
Class Gen (A : Type) := { arbitrary : G A }.
```

Generation Overview

```
Class Gen (A : Type) := { arbitrary : G A }.
```



Functions from
random seeds to A

The G Monad

```
returnGen : A -> G A.
```

```
bindGen : G A -> (A -> G B) -> G B.
```

The G Monad

```
returnGen : A -> G A.
```

Always return the
input argument, no
randomness

```
bindGen : G A -> (A -> G B) -> G B.
```

The G Monad

returnGen : A -> G A.

Always return the
input argument, no
randomness

And a way to
generate B given A...

bindGen : G A -> (A -> G B) -> G B.

Given generator for
A...

The G Monad

returnGen : A -> G A.

Always return the
input argument, no
randomness

And a way to
generate B given A...

bindGen : G A -> (A -> G B) -> G B.

Given generator for
A...

Produce a generator
for B

Primitive Generators - choose

```
choose : Choosable A -> A * A -> G A
```

Typeclass with
randomR

```
randomR : A * A -> RandomSeed -> A * RandomSeed
```

Primitive Generators - lists

```
listOf : G A -> G (list A)
```

```
vectorOf : nat -> G A -> G (list A)
```



Size of list to be generated

Generators - lists

How does listOf decide the size of the generated list?

```
listOf : G A -> G (list A)
```

```
vectorOf : nat -> G A -> G (list A)
```

Size of list to be generated

The G Monad – revisited

```
Inductive G A :=  
| MkG : (nat -> RandomSeed -> A) -> G A.
```


Custom Datatypes – Trees!

```
Inductive Tree A :=  
| Leaf : Tree A  
| Node : A -> Tree A -> Tree A -> Tree A.
```

Generator Combinator - oneOf

```
oneOf_ : G A -> list (G A) -> G A
```

Generator Combinator - oneOf

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```



Picks one at random

Generator Combinator - oneOf

```
oneOf_ : G A -> list (G A) -> G A
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Default element

Picks one at random

Generator Combinator - oneOf

```
oneOf_ : G A -> list (G A) -> G A
```

Default element

Picks one at random

```
Notation oneOf : list (G A) -> G A
```

A (naïve) random generator for trees

```
Fixpoint genTree: G tree :=  
  oneOf [ ret Leaf  
        , x <- arbitrary;  
          l <- genTree;  
          r <- genTree;  
          ret (Node x l r) ].
```

A (naïve) random generator

The type of tree generators

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Fixpoint genTree: G tree :=
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Uniform choice

A (naïve) random generator

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Point distribution
{Leaf}

Uniform choice

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Point distribution
{Leaf}

$x \in \text{Nat}$

$l \in \text{Tree}$

$r \in \text{Tree}$

Uniform choice

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Point distribution
{Leaf}

$x \in \text{Nat}$

$l \in \text{Tree}$

$r \in \text{Tree}$

Uniform choice

- Why does this terminate? (it doesn't)
- Is the distribution useful? (low probability of interesting trees)

A (naïve) random generator

The type of tree generators

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Fixpoint genTree: G tree :=
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Point distribution {Leaf}

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$l \in \text{Tree}$

$r \in \text{Tree}$

Uniform choice

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Leaf

A (naïve) random generator

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Point distribution
{Leaf}

$x \in \text{Nat}$

$l \in \text{Tree}$

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Uniform choice

- Why does this terminate? (it doesn't)
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Leaf
Leaf

A (naïve) random generator

The type of tree generators

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Point distribution {Leaf}

$x \in Nat$

$l \in Tree$

$r \in Tree$

Uniform choice

- Why does this terminate? (it doesn't)
- Is the distribution useful? (low probability of interesting trees)

Leaf
Leaf

Node 2 Leaf (Node 0 (Node 13 (Node 4 Leaf (Node 7 Leaf Leaf)) (Node 0 ...

A (better) random generator for t

size parameter :
upper limit of the
depth of the tree

```
Fixpoint genTree (size : nat) : G tree :=
```

$\{t \mid \text{size}(t) \leq \text{size}\}$

A (better) random generator for t

size parameter :
upper limit of the
depth of the tree

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if size = 0

```
  match size with
```

```
  | 0 => ret Leaf
```

if size =
size' + 1

```
  | S size' =>
```


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size' + 1

```
  | S size' =>
```

```
    oneOf [ ret Leaf
```

```
          , x <- arbitrary;
```

$x \in \text{Nat}$

```
          l <- genTree size';
```

$l \in \{t \mid \text{size}(t) \leq \text{size}'\}$

```
          r <- genTree size';
```

$r \in \{t \mid \text{size}(t) \leq \text{size}'\}$

```
          ret (Node x l r) ].
```

Recursive
calls with
smaller size

Generator Combinator - frequency

```
freq_ : G A -> list (nat * G A) -> G A
```

Default element

Picks one at random, using the weights!

```
Notation freq : list (nat * G A) -> G A
```

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  match size with  
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           , x <- arbitrary;  
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           ret (Node x l r) ].
```

$\{t \mid \text{size}(t) \leq \text{size}\}$

A (better) random generator for trees

size parameter :
upper limit of the
depth of the tree

```
Fixpoint genTree (size : nat) : G tree :=  
  match size with  
  | 0 => ret Leaf  
  | S size' =>  
    freq [ (1,      ret Leaf)  
          , (size, x <- arbitrary;  
            1 <- genTree size';  
            r <- genTree size';  
            ret (Node x 1 r)) ].
```

$\{t \mid \text{size}(t) \leq \text{size}\}$

$\frac{1}{\text{size}+1}$ of the time

$\frac{\text{size}}{\text{size}+1}$ of the time

Let's talk properties...

- Results
- Decidability
- Property combinators
- An example: tree mirroring

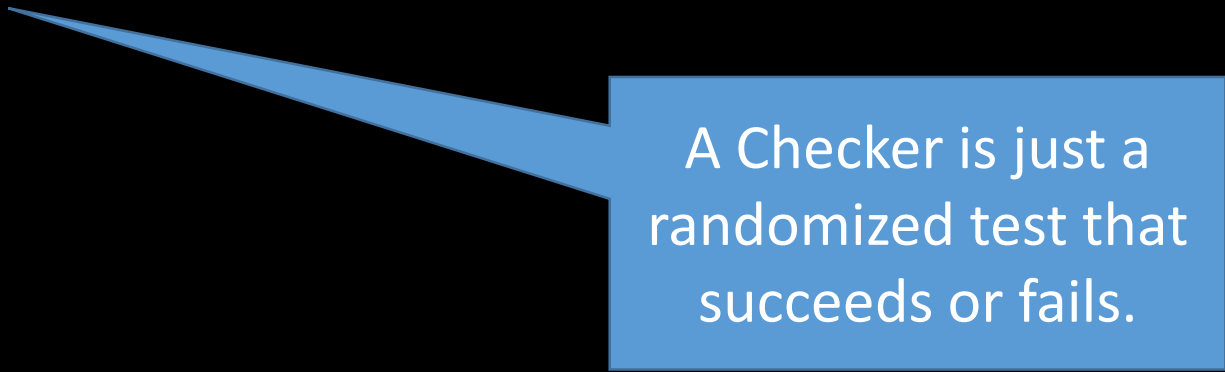
Results and Checkers

```
Inductive Result := Success | Failure.
```

Results and Checkers

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Definition Checker := G Result.



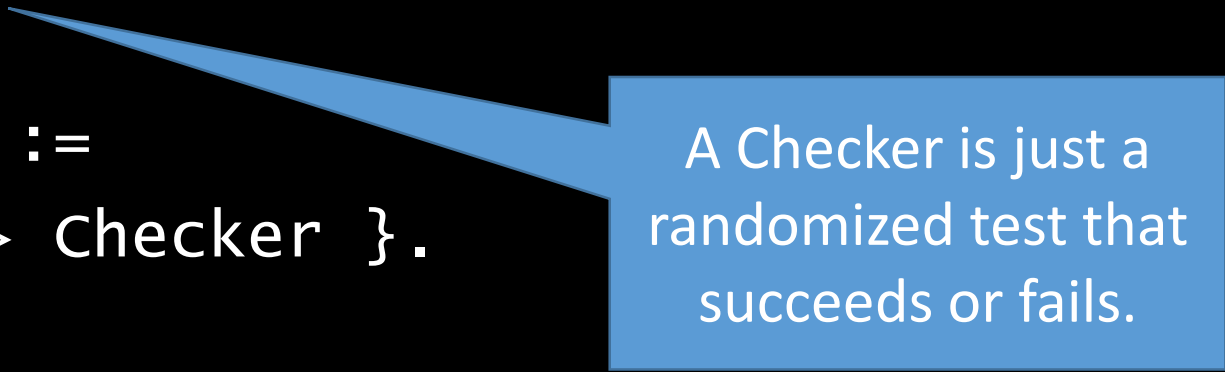
A Checker is just a randomized test that succeeds or fails.

Results and Checkers

Inductive Result := Success | Failure.

Definition Checker := G Result.

Class Checkable A :=
{ checker : A -> Checker }.



A Checker is just a randomized test that succeeds or fails.

Booleans are checkable

```
Instance CheckableBool : Checkable bool :=  
{ checker b :=  
  if b then ret Success else ret Failure  
}.  
.
```

Decidable Properties are Checkable

Class Dec P := { dec : {P} + {~P} }.

Notation P? := if dec P then true else false.

Decidable Properties are Checkable

```
Class Dec P := { dec : {P} + {~P} }.
```

```
Notation P? := if dec P then true else false.
```

```
Instance CheckDec {P} {Dec P} : Checkable P :=  
{ checker p := checker (P?) }.
```

Tree mirroring

```
Fixpoint mirror t : tree :=  
  match t with  
  | Leaf => Leaf  
  | Node x l r => Node x (mirror r) (mirror l)  
  end.
```

Tree mirroring

```
Fixpoint mirror t : tree :=  
  match t with  
  | Leaf => Leaf  
  | Node x l r => Node x (mirror r) (mirror l)  
end.
```

```
Definition mirrorP (t : tree nat) :=  
  mirror (mirror t) = t.
```

Putting it all together

```
Instance CheckFun {A P} {Gen A} {Checkable P}
: Checkable (A -> P) :=
{
  checker f :=
    a <- arbitrary ;
    checker (f a)
}.
```

Shrinking

```
Class Shrink A := { shrink : A -> list A }.
```

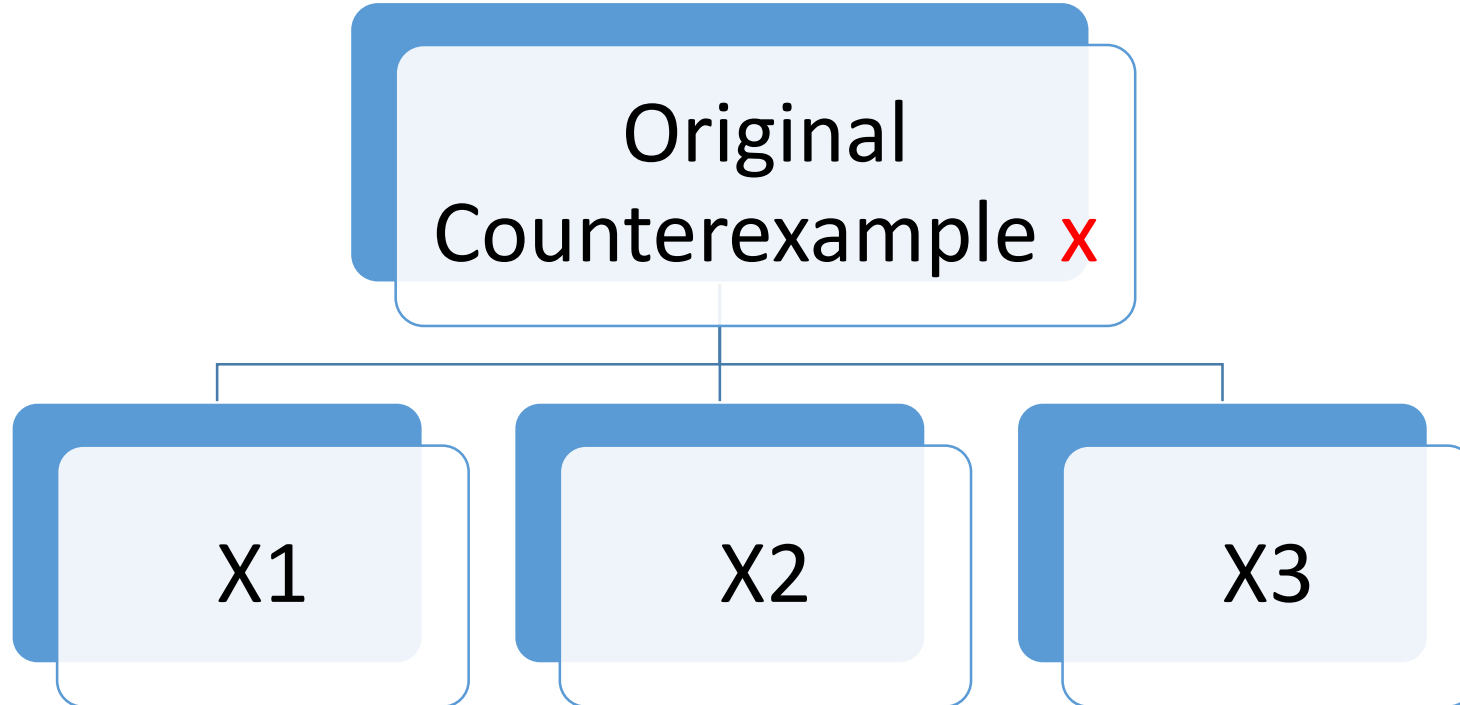
Shrinking

```
Class Shrink A := { shrink : A -> list A }.
```

Original
Counterexample **x**

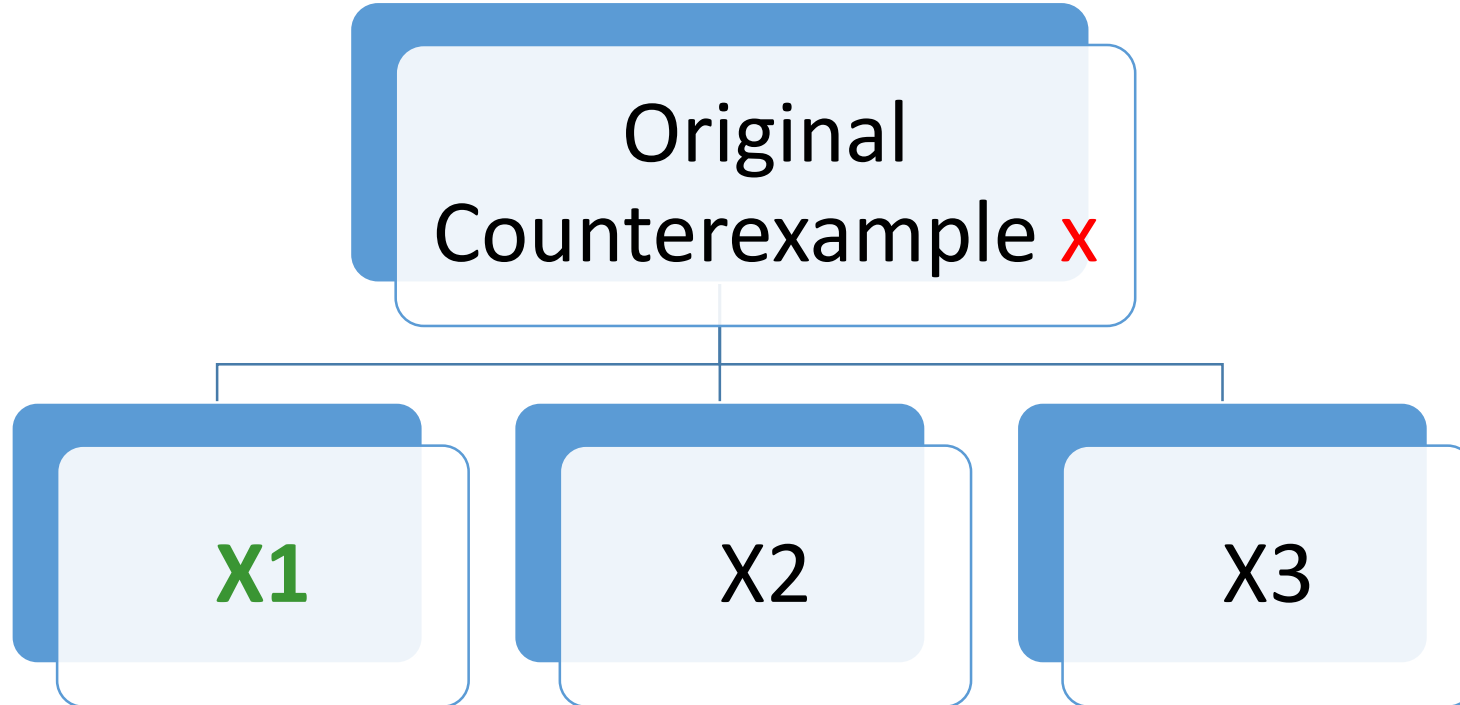
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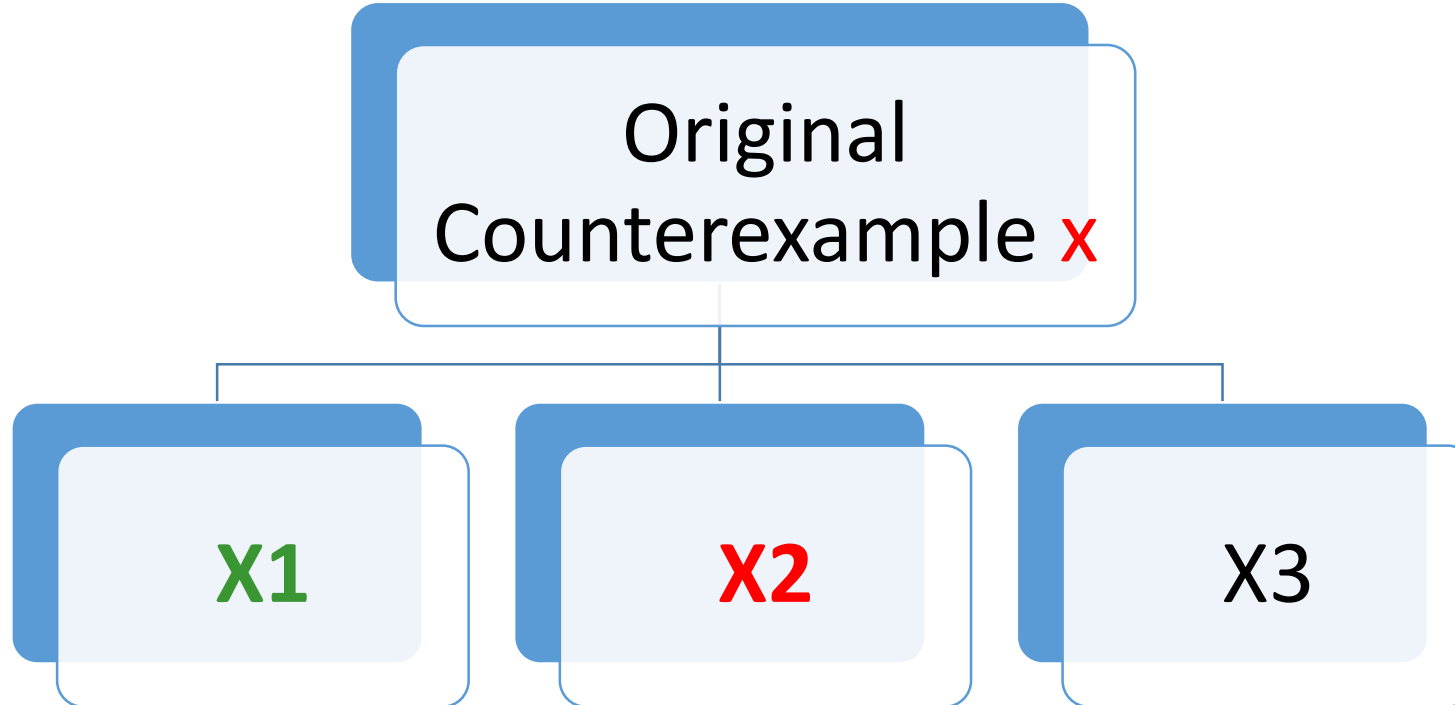
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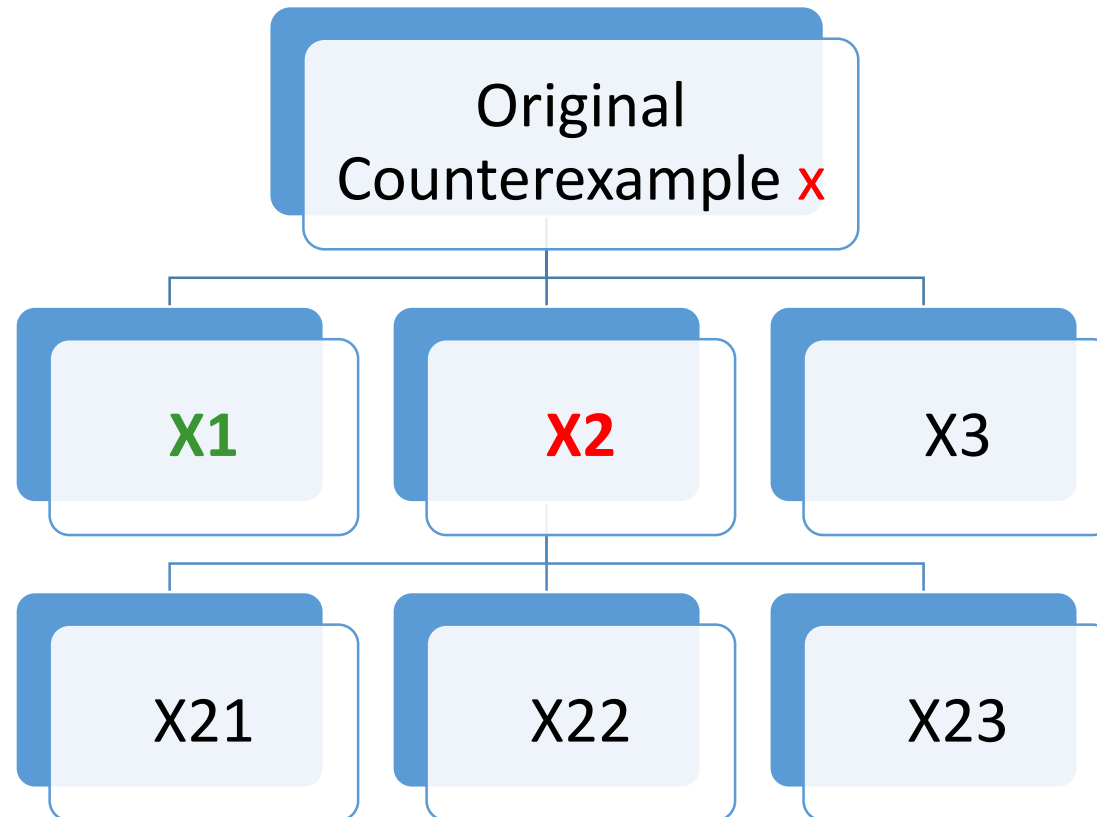
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Shrinking

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Class Shrink A := { shrink : A -> list A }.
```



Back to Preconditions!

$$\forall x. p(x) \rightarrow q(x)$$

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$$\forall x. p(x) \rightarrow q(x)$$

```
insert : A -> list A -> list A
```

```
sorted : list A -> bool
```

```
Definition insert_spec_sorted :=
```

```
  forall x l,
```


```
    sorted l -> sorted (insert x l).
```

Custom generators

Problem: Writing a good Generator

Custom generators

Problem: Writing a good Generator



All generated lists
are sorted

Custom generators

Problem: Writing a good Generator

All generated lists
are sorted

All sorted lists can
be generated

Custom generators

Problem: Writing a good Generator

All generated lists
are sorted

All sorted lists can
be generated

Distribution
appropriate for
testing

Take 2 – Custom Generators

Solution: Write a generator that produces well-type

Can be very complex
[ICFP 2013]

Problem: Writing a good generator

Problem: Too much boilerplate

Problem: Maintenance nightmare!

Testing feedback
should be
immediate

Generators and
predicates must be
kept in sync

Solution – Derive Generators Automatically

IDEA: Given a predicate p , produce a generator automatically

- Constrained Data with Uniform Distributions [FLOPS 2014]
- Making Random Judgments [ESOP 2015]
- Luck [POPL 2017]
- Generating Good Generators [POPL 2018]

Solution – Derive Generators Automatically

IDEA: Given a predicate p , produce a generator automatically

- Constrained Data with Uniform Distributions [FLOPS 2014]
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Problems:

- Addressing larger classes of preconditions
- Efficiency/Optimization
- Correctness
- Distribution <- big one!

More Open Problems!

- Deriving decidability procedures for inductive relations
- Shrinking, formalization and automation
- Connections with other areas
 - Probabilistic Programming
 - Fuzz Testing
 - Machine Learning

Thank you!

SOFTWARE FOUNDATIONS
VOLUME 4

QuickChick: Property-Based Testing in Coq

Leonidas Lampropoulos
Benjamin C. Pierce