



Lecture 11. Laws and induction

Functional Programming

**What does it mean for programs
to be equal/equivalent?**

Goals

- Equational reasoning: proving program equalities
- Reasoning principles at various types:
 - inductive proofs at algebraic data types;
 - extensional equality at function types.

Chapter 16 (up to 16.6) from Hutton's book

Laws

- Mathematical functions do not depend on hidden, changeable values
 - $2 + 3 = 5$, both in $4 \times (2 + 3)$ and in $(2 + 3)^2$
- This allows us to more easily prove properties that operators and functions might have
 - These properties are called **laws**

Examples of laws for integers

+ commutes

$$x + y = y + x$$

× commutes

$$x \times y = y \times x$$

+ is associative

$$x + (y + z) = (x + y) + z$$

× distributes over +

$$x \times (y + z) = x \times y + x \times z$$

0 is the unit of +

$$x + 0 = x = 0 + x$$

1 is the unit of ×

$$x \times 1 = x = 1 \times x$$

Why care about program equivalences?

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- Mathematical laws can help improve **performance**
 - That two expressions always have the same value does not mean that computing their value takes the same amount of time or memory
 - Replace a more expensive version with one that is cheaper to compute

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In short, performance and correctness

Equational reasoning by example

$$\begin{aligned} & (a + b)^2 \\ &= \text{-- definition of square} \\ & (a + b) \times (a + b) \\ &= \text{-- distributivity} \\ & ((a + b) \times a) + ((a + b) \times b) \\ &= \text{-- commutativity of } \times \\ & (a \times (a + b)) + (b \times (a + b)) \\ &= \text{-- distributivity, twice} \\ &= (a \times a + a \times b) + (b \times a + b \times b) \\ &= \text{-- associativity of } + \\ & a \times a + (a \times b + b \times a) + b \times b \\ &= \text{-- commutativity of } \times \\ & a \times a + (a \times b + a \times b) + b \times b \\ &= \text{-- definition of square and } (2 \times) \\ & a^2 + 2 \times a \times b + b^2 \end{aligned}$$

Each theory has its laws

- We have seen laws that deal with arithmetic operators
- During courses in logic you have seen similar laws for logic operators

commutativity of \wedge associativity of \wedge

$$x \wedge y = y \wedge x \quad x \wedge (y \wedge z) = (x \wedge y) \wedge z$$

distributivity of \wedge over \vee De Morgan's

$$x \wedge (y \vee z) = (x \wedge y) \vee (x \wedge z)$$

law Howard's law

$$\neg(x \wedge y) = \neg x \vee \neg y \quad (x \wedge y) \rightarrow z = x \rightarrow (y \rightarrow z)$$

A small proof in logic

$$\neg((a \vee b) \vee c) \rightarrow \neg d$$

$$= \text{-- De Morgan's law}$$

$$(\neg(a \vee b) \wedge \neg c) \rightarrow \neg d$$

$$= \text{-- De Morgan's law}$$

$$((\neg a \wedge \neg b) \wedge \neg c) \rightarrow \neg d$$

$$= \text{-- Howard's law}$$

$$(\neg a \wedge \neg b) \rightarrow (\neg c \rightarrow \neg d)$$

$$= \text{-- Howard's law}$$

$$\neg a \rightarrow (\neg b \rightarrow (\neg c \rightarrow \neg d))$$

- Proofs feel mechanical
 - You apply the “rules” implicit in the laws
 - Possibly even without understanding what \wedge and \vee do
- Always provide a hint why each equivalence holds!

- Haskell is referentially transparent
 - Calling a function twice with the same parameter is guaranteed to give the same result

Back to Haskell

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 - Calling a function twice with the same parameter is guaranteed to give the same result
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 - And use these to improve performance

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- Any = definition can be viewed in two ways

`double x = x + x`

1. The *definition* of a function
2. A *property* that can be used when reasoning
 - Replace `double x` by `x + x` and viceversa, for any `x`

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1. The *definition* of a function
 2. A *property* that can be used when reasoning
 - Replace `double x` by `x + x` and viceversa, for any `x`
- NB: by contrast, `<-` “assignments” in do-blocks are *not* referentially transparent!

A first example

For all compatible functions f and g , and lists xs

$$(\text{map } f \ . \ \text{map } g) \ xs = \text{map } (f \ . \ g) \ xs$$

This is not a definition, but a property/law

- The law can be shown to hold for the usual definitions of `map` and `(.)`

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Why care about this law?

The right-hand side is more performant than the left-hand side, in general

- Two traversals are combined into one

Relation to imperative languages

The law $\text{map } (f \ . \ g) = \text{map } f \ . \ \text{map } g$ is similar to the merging of subsequent loops

```
foreach (var elt in list) { stats1 }
```

```
foreach (var elt in list) { stats2 }
```

=

```
foreach (var elt in list) { stats1 ; stats2 }
```

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```
foreach (var elt in list) { stats2 }
```

=

```
foreach (var elt in list) { stats1 ; stats2 }
```

Due to side-effects in these languages, you have to be **really** careful when to apply them

- What could prevent us from merging the loops?

A few important laws

1. Function composition is associative

$$f \cdot (g \cdot h) = (f \cdot g) \cdot h$$

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- Validates executing a large map on different cores
- There is a generalization to lists of lists

$$\text{map } f \cdot \text{concat} = \text{concat} \cdot \text{map (map } f)$$

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- Validates executing a large map on different cores
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$$\text{map } f \cdot \text{concat} = \text{concat} \cdot \text{map (map } f)$$

3. `map` distributes over composition

$$\text{map (f \cdot g)} = \text{map } f \cdot \text{map } g$$

A few (more) important laws

4. If op is associative and e is the unit of op , then for finite lists xs

$$\text{foldr } op \ e \ xs = \text{foldl } op \ e \ xs$$

A few (more) important laws

4. If `op` is associative and `e` is the unit of `op`, then for finite lists `xs`

$$\text{foldr } op \ e \ xs = \text{foldl } op \ e \ xs$$

5. Under the same conditions, `foldr` on a singleton list is the identity

$$\text{foldr } op \ e \ [x] = x$$

These rules apply to very general functions

- The compiler uses these laws heavily to optimize

Why prove the laws?

- A proof guarantees that your optimization is justified
 - Otherwise you may accidentally change the behavior
- Proving is one additional way of increasing your confidence in the optimization that you perform
 - Others are testing, intuition, explanations...
- Of course, proofs can be wrong too
 - Proofs *can* be mechanically checked

Proving is like programming

1. Proposition = functionality of specification
2. Proof = implementation
3. Lemmas = library functions, local definitions

Proving is like programming

1. Proposition = functionality of specification
2. Proof = implementation
3. Lemmas = library functions, local definitions
4. Proof strategies = paradigms, design patterns
 - **Equational reasoning**, i.e., by a chain of equalities
 - **Proof by induction**
 - Proof by contraposition: prove p implies q by showing $\text{not } q$ implies $\text{not } p$
 - Proof by contradiction: assuming the opposite, show that leads to contradiction
 - Breaking down equalities: $x = y$ iff $x \leq y$ and $y \leq x$
 - Combinatorial proofs

Like programming, proving takes *practice*

Equational reasoning

foldr over a singleton list

If e is the unit element of op , then $\text{foldr } op \ e \ [x] = x$

`foldr` $op \ e \ [x]$

`= ...`

foldr over a singleton list

If e is the unit element of op , then $foldr\ op\ e\ [x] = x$

```
foldr op e [x]
```

```
= -- rewrite list notation
```

```
foldr op e (x : [])
```

```
= -- definition of foldr, case cons
```

```
op x (foldr op e [])
```

```
= -- definition of foldr, case empty
```

```
op x e
```

```
= -- e is neutral for op
```

```
x
```

foldl over a singleton list

If e is the unit element of op , then $foldl\ op\ e\ [x] = x$

```
foldl op e [x]
```

```
= ...
```

Try it yourself!

foldl over a singleton list

If e is the unit element of op , then $foldl\ op\ e\ [x] = x$

```
foldl op e [x]
```

```
= -- rewrite list syntactic sugar
```

```
foldl op e (x:[])
```

```
= -- definition foldl
```

```
foldl op (op e x) []
```

```
= -- definition foldl
```

```
op e x
```

```
= -- e is neutral for op
```

```
x
```

Function composition is associative

For all functions f, g and h , $f \circ (g \circ h) = (f \circ g) \circ h$

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For all functions f, g and h , $f \circ (g \circ h) = (f \circ g) \circ h$

Proof: consider any x

$$\begin{aligned} & (f \circ (g \circ h)) x \\ &= \text{-- definition of } (\circ) \\ & f ((g \circ h) x) \\ &= \text{-- definition of } (\circ) \\ & f (g (h x)) \\ &= \text{-- definition of } (\circ) \\ & (f \circ g) (h x) \\ &= \text{-- definition of } (\circ) \\ & ((f \circ g) \circ h) x \end{aligned}$$

Proving functions equal

- We prove functions f and g equal by proving that for all input x , $f(x) = g(x)$
 - They give the same results for the same inputs
 - Provided that they don't have side effects!
- They need *not* be the same function, as long as they behave in the same way
 - We call this **extensional** equality
- It is essential to make *no* assumptions about x
 - Otherwise, the proof does not work *for all* x

Two column style proofs

Reasoning from two ends is typically easier

- Rewrite the expression until you reach the same point
- Equalities can be read “backwards”

For all functions f, g and h , $f \circ (g \circ h) = (f \circ g) \circ h$

Proof: consider any x

$(f \circ (g \circ h)) x$	$((f \circ g) \circ h) x$
$= \{- \text{ defn. of } (\circ) -\}$	$= \{- \text{ defn. of } (\circ) -\}$
$f ((g \circ h) x)$	$(f \circ g) (h x)$
$= \{- \text{ defn. of } (\circ) -\}$	$= \{- \text{ defn. of } (\circ) -\}$
$f (g (h x))$	$f (g (h x))$

map after (:)

For all type compatible values x and functions f ,

$$\text{map } f \ . \ (x \ :) = (f \ x \ :) \ . \ \text{map } f$$

map after (:)

For all type compatible values x and functions f ,

`map f . (x :)` = `(f x :)` . `map f`

Proof: consider any list xs

<code>(map f . (x :)) xs</code>	<code>((f x :) . map f) xs</code>
<code>= {- defn of (.) -}</code>	<code>= {- defn of (.) -}</code>
<code>map f ((x :) xs)</code>	<code>(f x :) (map f xs)</code>
<code>= {- section notation -}</code>	<code>= {- section notation -}</code>
<code>map f (x : xs)</code>	<code>f x : map f xs</code>
<code>= {- defn. of map -}</code>	
<code>f x : map f xs</code>	

not is an involution

The functions `not . not` and `id` are equal

Let's try!

not is an involution

The functions `not . not` and `id` are equal

Proof: consider any Boolean value `x`

- Case `x = False`

<code>(not . not) False</code>	<code>id False</code>
<code>= {- defn of (.) -}</code>	<code>= {- defn. of id -}</code>
<code>not (not False)</code>	<code>False</code>
<code>= {- defn of not -}</code>	
<code>not True</code>	
<code>= {- defn of not -}</code>	
<code>False</code>	

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<code>= {- defn of not -}</code>	
<code>not True</code>	
<code>= {- defn of not -}</code>	
<code>False</code>	

- Case `x = True`

<code>(not . not) True</code>	<code>id True</code>
<code>= {- as above -}</code>	<code>= {- defn. of id -}</code>
<code>True</code>	<code>True</code>

Case distinction

- To prove a property *for all* x , sometimes we need to distinguish the possible shapes that x may take
 - We need to be exhaustive to cover *all* cases

Case distinction

- To prove a property *for all* x , sometimes we need to distinguish the possible shapes that x may take
 - We need to be exhaustive to cover *all* cases
- For example,
 - A Boolean may be either True or False
 - A Maybe a value could be Nothing or Just x for some x
 - Given a data type of the form

```
data Shape = Circle    Point Float
           | Rectangle Point Float Float
           | Triangle  Point Point Point
```

you need to consider three different cases

Homework: Booleans and (&&) form a monoid

1. True is a neutral element: for any Boolean x ,

$$\text{True} \ \&\& \ x = x$$

$$x \ \&\& \ \text{True} = x$$

2. (&&) is associative: for any Booleans x , y , and z ,

$$x \ \&\& \ (y \ \&\& \ z) = (x \ \&\& \ y) \ \&\& \ z$$

Homework: Maybe a forms a monoid

Consider the following operation:

`Just x <|> _ = Just x`

`Nothing <|> y = y`

1. `Nothing` is a neutral element: for any `x :: Maybe a`,

`Nothing <|> x = x`

`x <|> Nothing = x`

2. `(<|>)` is associative

Induction on data types

The case for lists

- Every (finite) list is built by finitely many $(:)$ 'es applied to a final $[]$
 $x : (y : (z : \dots (w : [])))$
 - Don't bother about (finite) for now
- What if ...?
 - we prove a property P for $[]$
 - given any list xs satisfying P , we can prove P holds for $x : xs$
- The *(structural) induction principle for (finite) lists* says that the result then holds **for all** finite lists

The case for numbers and trees

- Every finite natural number can be seen as applying the successor function finitely many times to 0

$$4 = \text{Succ} (\text{Succ} (\text{Succ} (\text{Succ} \text{Zero})))$$

- What if...?
 - we prove a property P for 0
 - given a number n satisfying P , we can prove P for $\text{succ } n = n + 1$

The case for numbers and trees

- Every finite natural number can be seen as applying the successor function finitely many times to 0

$$4 = \text{Succ} (\text{Succ} (\text{Succ} (\text{Succ} \text{Zero})))$$

- What if...?
 - we prove a property P for 0
 - given a number n satisfying P , we can prove P for $\text{succ } n = n + 1$
- Every (finite) binary tree is built by finitely many Nodes ultimately applied to Leaf
 - What if...?
 - we prove a property P for Leaf
 - given any two trees l and r satisfying P and a value x , we can prove P for $\text{Node } l \ x \ r$

A strategy for proving properties of structured data

1. State the law
 - a. If we speak about functions, introduce input variables
2. Enumerate the cases for one of the variables
 - Usually, one per constructor in the data type
3. Prove the base cases by equational reasoning
4. Prove the recursive cases
 - a. State the *induction hypotheses* (IH)
 - b. Use equational reasoning, applying IH when needed

1. State the law
 - a. If we speak about functions, introduce input variables
 - b. If needed, choose a variable to perform induction on
2. Prove the case [] by equational reasoning
3. State the induction hypothesis for xs
4. Prove the case x : xs, assuming that the IH holds

map f distributes over (++)

For all lists xs and ys

`map f (xs ++ ys) = map f xs ++ map f ys`

map f distributes over (++)

For all lists xs and ys

`map f (xs ++ ys) = map f xs ++ map f ys`

Proof: by induction on xs

- Case xs = []

<code>map f ([] ++ ys)</code>	<code>map f [] ++ map f ys</code>
<code>= {- defn. of (++) -}</code>	<code>= {- defn. of map -}</code>
<code>map f ys</code>	<code>[] ++ map f ys</code>
	<code>= {- defn of (++) -}</code>
	<code>map f ys</code>

map f distributes over (++)

- Case $xs = z:zs$
 - IH: $\text{map } f (zs ++ ys) = \text{map } f zs ++ \text{map } f ys$

<code>map f ((z:zs) ++ ys)</code>	<code>map f (z:zs) ++ map f ys</code>
<code>= {- defn. of (++) -}</code>	<code>= {- defn. of map -}</code>
<code>map f (z : (zs ++ ys))</code>	<code>(f z : map f zs) ++ map f ys</code>
<code>= {- defn of map -}</code>	<code>= {- defn of (++) -}</code>
<code>f z : map f (zs ++ ys)</code>	<code>f z : (map f zs ++ map f ys)</code>
	<code>= {- IH -}</code>
	<code>f z : map f (zs ++ ys)</code>

map distributes over composition

For all compatible functions f and g ,

$$\text{map } (f \ . \ g) = \text{map } f \ . \ \text{map } g$$

Proof: by extensionality, we need to prove that for all xs

$$\text{map } (f \ . \ g) \ xs = (\text{map } f \ . \ \text{map } g) \ xs$$

map distributes over composition

For all compatible functions f and g ,

$$\text{map } (f \ . \ g) = \text{map } f \ . \ \text{map } g$$

Proof: by extensionality, we need to prove that for all xs

$$\text{map } (f \ . \ g) \ xs = (\text{map } f \ . \ \text{map } g) \ xs$$

We proceed by induction on xs

- Case $xs = []$

$$\begin{aligned} \text{map } (f \ . \ g) \ [] & \quad (\text{map } f \ . \ \text{map } g) \ [] \\ = \{- \text{ defn. of map -} \} & \quad = \{- \text{ defn of } (.) \ -\} \\ [] & \quad \text{map } f \ (\text{map } g \ []) \\ & \quad = \{- \text{ defn. of map, twice -} \} \\ & \quad [] \end{aligned}$$

map distributes over composition

- Case $xs = z:zs$
 - IH: $\text{map } (f \ . \ g) \ zs = (\text{map } f \ . \ \text{map } g) \ zs$

<code>map (f.g) (z:zs)</code>	<code>(map f . map g) (z:zs)</code>
<code>= {- defn. of map -}</code>	<code>= {- defn. of (.) -}</code>
<code>(f.g) z : map (f.g) zs</code>	<code>map f (map g (z:zs))</code>
<code>= {- defn of (.) -}</code>	<code>= {- defn. of map -}</code>
<code>f (g z) : map (f.g) zs</code>	<code>map f (g z : map g zs)</code>
	<code>= {- defn. of map -}</code>
	<code>f (g z) : map f (map g zs)</code>
	<code>= {- IH -}</code>
	<code>f (g z) : map (f.g) zs</code>

reverse is an involution

The functions `reverse . reverse` and `id` are equal

Proof: by extensionality we need to prove that for all `xs`

```
(reverse . reverse) xs  
= reverse (reverse xs)    = id xs
```

reverse is an involution

The functions `reverse . reverse` and `id` are equal

Proof: by extensionality we need to prove that for all `xs`

```
(reverse . reverse) xs
= reverse (reverse xs) = id xs
```

We proceed by induction on `xs`

- Case `xs = []`

```
reverse (reverse [])      id []
= {- defn. of reverse -}  = {- defn. of id -}
reverse []                []
= {- defn. of reverse -}
[]
```

reverse is an involution

- Case $xs = z:zs$

- IH: $\text{reverse} (\text{reverse } zs) = \text{id } zs = zs$

```
reverse (reverse (z:zs))    id (z:zs)
= {- defn. of reverse -}    = {- defn of id -}
reverse (reverse zs ++ [z]) z:zs
```

We are stuck!

Lemmas

To keep going we defer some parts as *lemmas*

- Similar to local definitions in code
- Lemmas have to be proven separately

In our case, we need the following lemmas

```
-- Distributivity of (++) over reverse
reverse (xs ++ ys) = reverse ys ++ reverse xs
-- Reverse on singleton lists
reverse [x]          = [x]
```

Finding the right lemmas involves lots of practice

reverse is an involution

```
reverse (reverse (z:zs))
= {- defn. of reverse -}
reverse (reverse zs ++ [z])
= {- distributivity -}
reverse [z] ++ reverse (reverse zs)
= {- reverse on singleton -}
[z] ++ reverse (reverse zs)
= {- IH -}
[z] ++ zs                                id (z : zs)
= {- defn of (++) -}                      = {- defn of id -}
z : zs                                    z : zs
```

We still need to prove the lemmas separately

reverse is an involution

Lemma: `reverse (xs++ys) = reverse ys ++ reverse xs`

Proof: by induction on `xs` ...

Lemma: `reverse [x] = [x]`

Proof:

`reverse [x]`

`= {- list notation -}`

`reverse (x : [])`

`= {- defn. of reverse -}`

`reverse [] ++ [x]`

`= {- defn. of reverse -}`

`[] ++ [x]`

`= {- defn. of (++) -}`

`[x]`

Mathematical induction

- To prove that a statement P holds for all $n \in \mathbb{N}$
 - Prove that it holds for 0
 - Prove that it holds for $n + 1$ assuming that it holds for n
- This strategy is equivalent to structural induction on

data Nat = Zero | Succ Nat

This encoding is called *Peano numbers*

Note: there are stronger forms of induction for natural numbers, but we restrict ourselves to the simpler one

Arithmetic using Peano numbers

Addition and multiplication are defined by recursion

```
add  :: Nat -> Nat -> Nat
```

```
add  Zero      m = m
```

```
--      0 + m = m
```

```
add  (Succ n) m = Succ (n + m)
```

```
-- (n + 1) + m = (n + m) + 1
```

```
mult :: Nat -> Nat -> Nat
```

```
mult Zero      m = Zero
```

```
--      0 × m = 0
```

```
mult (Succ n) m = add (mult n m) m
```

```
-- (n + 1) × m = (n × m) + m
```

0 is right identity for addition

For all natural n , $\text{add } n \text{ Zero} = n$

Proof: by induction on n

- Case $n = \text{Zero}$

$\text{add } \text{Zero } \text{Zero}$

$= \{- \text{ defn. of add } -\}$

Zero

- Case $n = \text{Succ } p$

- IH: $\text{add } p \text{ Zero} = p$

$\text{add } (\text{Succ } p) \text{ Zero}$

$= \{- \text{ defn. of add } -\}$

$\text{Succ } (\text{add } p \text{ Zero})$

$= \{- \text{ IH } -\}$

$\text{Succ } p$

Some functions over binary trees

```
data Tree a = Leaf | Node (Tree a) a (Tree a)
```

size t counts the number of nodes

```
size Leaf = 0
```

```
size (Node l _ r) = 1 + size l + size r
```

mirror t obtains the “rotated” image of a tree

```
mirror Leaf = Leaf
```

```
mirror (Node l x r) = Node (mirror r) x (mirror l)
```

mirror preserves the size

For all trees t , $\text{size} (\text{mirror } t) = \text{size } t$

mirror preserves the size

For all trees t , $\text{size} (\text{mirror } t) = \text{size } t$

Proof: by induction on t

- Case $t = \text{Leaf}$

```
size (mirror Leaf)
= {- defn. of mirror -}
size Leaf
```

mirror preserves the size

- Case $t = \text{Node } l \times r$
 - We get one induction hypothesis per recursive position
 - IH1: $\text{size } (\text{mirror } l) = \text{size } l$
 - IH2: $\text{size } (\text{mirror } r) = \text{size } r$

```
size (mirror (Node l x r))
= {- defn. of mirror -}
size (Node (mirror r) x (mirror l))
= {- defn. of size -}
1 + size (mirror r) + size (mirror l)
= {- IH1 and IH2 -}
1 + size r + size l
= {- commutativity of addition -}
1 + size l + size r
= {- defn. of size -}
size (Node l x r)
```

0 is an absorbing element for product

For all natural n , $\text{mult } n \text{ Zero} = \text{Zero}$

- Proving program equivalences is useful for
 - establishing *correctness*;
 - finding opportunities for improving *performance*;
- We prove equivalences using
 - *definitions* and *laws*;
 - *extensional equality* at function types;
 - *case distinction* and *induction* on algebraic data types;

Some advice

- Proving takes practice, just like programming
 - *So practice*
 - Both the book and the lecture notes contain many more examples of inductive proofs
- Inductive proofs are *definitely* part of the final exam
 - Could be about lists, natural numbers, trees, or some other recursively defined data type