PROGRAMMING LANGUAGES: FUNCTIONAL PROGRAMMING

3. DEFINITION AND PROOF BY INDUCTION

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TOTAL FUNCTIONAL PROGRAMMING

- The next few lectures concerns inductive definitions and proofs of datatypes and programs.
- While Haskell provides allows one to define nonterminating functions, infinite data structures, for now we will only consider its total, finite fragment.
- That is, we temporarily
 - consider only finite data structures,
 - demand that functions terminate for all value in its input type, and
 - provide guidelines to construct such functions.
- Infinite datatypes and non-termination will be discussed later in this course.

INDUCTION ON NATURAL NUMBERS

THE SO-CALLED "MATHEMATICAL INDUCTION"

- Let *P* be a predicate on natural numbers.
- We've all learnt this principle of proof by induction: to prove that *P* holds for all natural numbers, it is sufficient to show that
 - P0 holds;
 - P(1+n) holds provided that Pn does.

PROOF BY INDUCTION ON NATURAL NUMBERS

- We can see the above inductive principle as a result of seeing natural numbers as defined by the datatype ¹ data Nat = $0 | \mathbf{1}_+ Nat$.
- That is, any natural number is either 0, or **1**₊ *n* where *n* is a natural number.
- In this lecture, 1₊ is written in bold font to emphasise that it is a data constructor (as opposed to the function (+), to be defined later, applied to a number 1).

¹Not a real Haskell definition.

A PROOF GENERATOR

Given P0 and $Pn \Rightarrow P(1+n)$, how does one prove, for example, P 3?

 $P (1_{+} (1_{+} (1_{+} 0))) \\ \Leftarrow \{ P (1_{+} n) \Leftrightarrow Pn \} \\ P (1_{+} (1_{+} 0)) \\ \Leftrightarrow \{ P (1_{+} n) \Leftrightarrow Pn \} \\ P (1_{+} 0) \\ \Leftrightarrow \{ P (1_{+} n) \Leftrightarrow Pn \} \\ P0 .$

Having done math. induction can be seen as having designed *a program that generates a proof* — given any *n* :: *Nat* we can generate a proof of *P n* in the manner above.

INDUCTIVELY DEFINED FUNCTIONS

• Since the type *Nat* is defined by two cases, it is natural to define functions on *Nat* following the structure:

 $\begin{array}{ll} exp & :: \operatorname{Nat} \to \operatorname{Nat} \to \operatorname{Nat} \\ exp \ b \ 0 & = 1 \\ exp \ b \ (\mathbf{1}_{+} \ n) & = b \times exp \ b \ n \ . \end{array}$

• Even addition can be defined inductively

 $\begin{array}{ll} (+) & :: \operatorname{Nat} \to \operatorname{Nat} \to \operatorname{Nat} \\ 0+n & = n \\ (\mathbf{1}_+ \ m) + n = \mathbf{1}_+ \ (m+n) \end{array}.$

• Exercise: define (×)?

A VALUE GENERATOR

Given the definition of *exp*, how does one compute *exp b* 3?

exp b (**1**₊ (**1**₊ (**1**₊ 0)))

- = { definition of exp } $b \times exp \ b (1_+ (1_+ 0))$
- = { definition of exp } $b \times b \times exp \ b (\mathbf{1}_{+} \ \mathbf{0})$
- = { definition of exp }
 - $b \times b \times b \times exp \ b \ 0$
- = { definition of *exp* }

 $b \times b \times b \times 1$.

It is a program that generates a value, for any *n* :: *Nat*. Compare with the proof of *P* above.

An inductive proof is a program that generates a proof for any given natural number.

An inductive program follows the same structure of an inductive proof.

Proving and programming are very similar activities.

WITHOUT THE n + k PATTERN

• Unfortunately, newer versions of Haskell abandoned the "n + k pattern" used in the previous slides:

```
\begin{array}{ll} exp & :: \ Int \rightarrow Int \rightarrow Int \\ exp \ b \ 0 &= 1 \\ exp \ b \ n &= b \times exp \ b \ (n-1) \end{array} .
```

- *Nat* is defined to be *Int* in MiniPrelude.hs. Without MiniPrelude.hs you should use *Int*.
- For the purpose of this course, the pattern 1 + n reveals the correspondence between Nat and lists, and matches our proof style. Thus we will use it in the lecture.
- Remember to remove them in your code.

- To prove properties about *Nat*, we follow the structure as well.
- E.g. to prove that $exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n$.
- One possibility is to preform induction on *m*. That is, prove Pm for all m :: Nat, where $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$

Recall $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$ Case m := 0. For all n, we reason: $exp \ b \ (0+n)$

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exp \ b \ (0+n)= \begin{cases} defn. of (+) \\ exp \ b \ n \end{cases}
```

```
Recall Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).

Case m := 0. For all n, we reason:

exp \ b \ (0+n)
= \{ defn. of (+) \}
exp \ b \ n
= \{ defn. of (\times) \}
1 \times exp \ b \ n
```

```
Recall Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).
Case m := 0. For all n, we reason:
          exp b (0+n)
      = \{ defn. of (+) \}
          exp b n
      = { defn. of (x) }
          1 \times exp b n
      = \{ defn. of exp \}
          exp b 0 \times exp b n.
```

We have thus proved P0.

Recall $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$ Case $m := \mathbf{1}_+ m$. For all n, we reason: $exp \ b \ ((\mathbf{1}_+ m) + n)$

Recall $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$ Case $m := \mathbf{1}_+ m$. For all n, we reason: $exp \ b \ ((\mathbf{1}_+ m) + n)$ $= \begin{cases} exp \ b \ ((\mathbf{1}_+ m) + n) \\ defn. \ of \ (+) \end{cases}$ $exp \ b \ (\mathbf{1}_+ \ (m+n))$

Recall $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$ Case $m := \mathbf{1}_+ m$. For all n, we reason: $exp \ b \ ((\mathbf{1}_+ \ m) + n)$ $= \{ defn. of (+) \}$ $exp \ b \ (\mathbf{1}_+ \ (m+n))$ $= \{ defn. of \ exp \}$ $b \times exp \ b \ (m+n)$

Recall $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$ Case $m := \mathbf{1}_+ m$. For all n, we reason:

```
exp \ b \ ((1_+ \ m) + n)
= { defn. of (+) }
```

```
exp \ b \ (\mathbf{1}_{+} \ (m+n))
```

- = { defn. of exp }
 - $b \times exp \ b \ (m+n)$
- = { induction }

 $b \times (exp \ b \ m \times exp \ b \ n)$

Recall $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$ Case $m := \mathbf{1}_+ m$. For all n, we reason:

exp b ((1+m)+n) $= \{ defn. of (+) \}$ exp b (1+(m+n)) $= \{ defn. of exp \}$ $b \times exp b (m+n)$ $= \{ induction \}$ $b \times (exp b m \times exp b n)$ $= \{ (\times) associative \}$

 $(b \times exp \ b \ m) \times exp \ b \ n$

Recall $Pm \equiv (\forall n :: exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n).$ Case $m := \mathbf{1}_+ m$. For all *n*, we reason: exp b ((1 + m) + n) $= \{ defn. of (+) \}$ exp b $(1_{+}(m+n))$ $= \{ defn. of exp \}$ $b \times exp b (m+n)$ = { induction } $b \times (exp \ b \ m \times exp \ b \ n)$ $= \{ (\times) \text{ associative } \}$ $(b \times exp \ b \ m) \times exp \ b \ n$ = { defn. of exp } $exp b (1_+ m) \times exp b n$.

We have thus proved $P(1_+ m)$, given Pm.

- The inductive proof could be carried out smoothly, because both (+) and *exp* are defined inductively on its lefthand argument (of type *Nat*).
- The structure of the proof follows the structure of the program, which in turns follows the structure of the datatype the program is defined on.

LISTS AND NATURAL NUMBERS

- We have yet to prove that (\times) is associative.
- The proof is quite similar to the proof for associativity of (++), which we will talk about later.
- In fact, *Nat* and lists are closely related in structure.
- Most of us are used to think of numbers as atomic and lists as structured data. Neither is necessarily true.
- For the rest of the course we will demonstrate induction using lists, while taking the properties for *Nat* as given.

AN INDUCTIVELY DEFINED SET?

- For a set to be "inductively defined", we usually mean that it is the *smallest* fixed-point of some function.
- What does that maen?

FIXED-POINT AND PREFIXED-POINT

- A fixed-point of a function f is a value x such that fx = x.
- **Theorem**. *f* has fixed-point(s) if *f* is a *monotonic function* defined on a complete lattice.
 - In general, given *f* there may be more than one fixed-point.
- A prefixed-point of f is a value x such that $fx \le x$.
 - Apparently, all fixed-points are also prefixed-points.
- **Theorem**. the smallest prefixed-point is also the smallest fixed-point.

Example: Nat

- Recall the usual definition: *Nat* is defined by the following rules:
 - 1. 0 is in *Nat*;
 - 2. if *n* is in *Nat*, so is **1**₊ *n*;
 - 3. there is no other *Nat*.
- If we define a function *F* from sets to sets: $FX = \{0\} \cup \{\mathbf{1}_{+} \ n \mid n \in X\}, 1$. and 2. above means that $FNat \subseteq Nat$. That is, *Nat* is a prefixed-point of *F*.
- 3. means that we want the *smallest* such prefixed-point.
- Thus Nat is also the least (smallest) fixed-point of F.

Formally, let $FX = \{0\} \cup \{\mathbf{1}_+ \ n \mid n \in X\}$, Nat is a set such that $F \operatorname{Nat} \subseteq \operatorname{Nat} ,$ (1) $(\forall X : FX \subseteq X \Rightarrow \operatorname{Nat} \subseteq X) ,$ (2)

where (1) says that Nat is a prefixed-point of F, and (2) it is the least among all prefixed-points of F.

MATHEMATICAL INDUCTION, FORMALLY

- Given property *P*, we also denote by *P* the set of elements that satisfy *P*.
- That P0 and $Pn \Rightarrow P(\mathbf{1}_{+}n)$ is equivalent to $\{0\} \subseteq P$ and $\{\mathbf{1}_{+} n \mid n \in P\} \subseteq P$,
- which is equivalent to $FP \subseteq P$. That is, P is a prefixed-point!
- By (2) we have $Nat \subseteq P$. That is, all Nat satisfy P!
- This is "why mathematical induction is correct."

There is a dual technique called *coinduction* where, instead of least prefixed-points, we talk about *greatest postfixed points*. That is, largest x such that $x \le fx$.

With such construction we can talk about infinite data structures.

INDUCTION ON LISTS

• Recall that a (finite) list can be seen as a datatype defined by: ²

data List a = [] | a : List a.

• Every list is built from the base case [], with elements added by (:) one by one: [1, 2, 3] = 1 : (2 : (3 : [])).

²Not a real Haskell definition.

But what about infinite lists?

- For now let's consider finite lists only, as having infinite lists make the *semantics* much more complicated. ³
- In fact, all functions we talk about today are total functions. No \perp involved.

³What does that mean? We will talk about it later.

SET-THEORETICALLY SPEAKING...

The type List a is the smallest set such that

- 1. [] is in *List a*;
- 2. if xs is in List a and x is in a, x : xs is in List a as well.

INDUCTIVELY DEFINED FUNCTIONS ON LISTS

• Many functions on lists can be defined according to how a list is defined:

sum :: List $Int \rightarrow Int$ sum [] = 0 sum (x : xs) = x + sum xs .

 $\begin{array}{ll} map & :: (a \rightarrow b) \rightarrow List \ a \rightarrow List \ b \\ map \ f[] & = [] \\ map \ f(x:xs) = fx: map \ fxs \ . \end{array}$

• The function (++) appends two lists into one

 $(++) :: List a \rightarrow List a \rightarrow List a$ [] ++ ys = ys(x : xs) ++ ys = x : (xs ++ ys) .

• Compare the definition with that of (+)!

PROOF BY STRUCTURAL INDUCTION ON LISTS

- Recall that every finite list is built from the base case [], with elements added by (:) one by one.
- To prove that some property *P* holds for all finite lists, we show that
 - 1. *P* [] holds;
 - 2. forall x and xs, P(x : xs) holds provided that P xs holds.

FOR A PARTICULAR LIST...

Given P[] and $P xs \Rightarrow P(x : xs)$, for all x and xs, how does one prove, for example, P[1, 2, 3]?

$$P(1:2:3:[]) \\ \Leftarrow \ \{ P(x:xs) \Leftarrow Pxs \} \\ P(2:3:[]) \\ \Leftrightarrow \ \{ P(x:xs) \Leftarrow Pxs \} \\ P(3:[]) \\ \Leftarrow \ \{ P(x:xs) \Leftarrow Pxs \} \\ P[3:[]) \\ \Leftrightarrow \ \{ P(x:xs) \Leftarrow Pxs \} \\ P[].$$

APPENDING IS ASSOCIATIVE

To prove that xs + (ys + zs) = (xs + ys) + zs.

Let $P xs = (\forall ys, zs :: xs ++(ys ++ zs) = (xs ++ ys) ++ zs)$, we prove P by induction on xs.

Case xs := []. For all ys and zs, we reason:

[] ++ (ys ++ zs) = { defn. of (++) }

ys ++ zs

= { defn. of (++) } ([]++ys)++zs .

We have thus proved P [].

APPENDING IS ASSOCIATIVE

Case xs := x : xs. For all ys and zs, we reason: (x : xs) ++ (ys ++ zs) $= \{ defn. of (++) \}$ x : (xs ++(ys ++ zs))= { induction } x: ((xs ++ ys) ++ zs) $= \{ defn. of (++) \}$ (x : (xs ++ ys)) ++ zs $= \{ defn. of (++) \}$ ((x : xs) ++ vs) ++ zs.

We have thus proved P(x : xs), given P xs.

Do We Have To Be So Formal?

- In our style of proof, every step is given a reason. Do we need to be so pedantic?
- Being formal *helps* you to do the proof:
 - In the proof of $exp \ b \ (m+n) = exp \ b \ m \times exp \ b \ n$, we expect that we will use induction to somewhere. Therefore the first part of the proof is to generate $exp \ b \ (m+n)$.
 - In the proof of associativity, we were working toward generating xs ++(ys ++ zs).
- By being formal we can work on the *form*, not the *meaning*. Like how we solved the knight/knave problem
- Being formal actually makes the proof easier!
- Make the symbols do the work.

LENGTH

• The function *length* defined inductively:

 $\begin{array}{ll} \mbox{length} & :: \mbox{List } a \to Nat \\ \mbox{length} \left[\right] & = 0 \\ \mbox{length} \left(x : xs \right) = {\bf 1}_+ (\mbox{length} xs) \ . \end{array}$

• Exercise: prove that *length* distributes into (++):

length (xs ++ ys) = length xs + length ys

• While (++) repeatedly applies (:), the function *concat* repeatedly calls (++):

concat	:: List (List a) \rightarrow List a
concat []	=[]
concat (xs : xss)	= xs ++ concat xss .

- Compare with sum.
- Exercise: prove $sum \cdot concat = sum \cdot map sum$.

DEFINITION BY INDUCTION/RECURSION

- Rather than giving commands, in functional programming we specify values; instead of performing repeated actions, we define values on inductively defined structures.
- Thus induction (or in general, recursion) is the only "control structure" we have. (We do identify and abstract over plenty of patterns of recursion, though.)
- To inductively define a function f on lists, we specify a value for the base case (f []) and, assuming that f xs has been computed, consider how to construct f (x : xs) out of f xs.

• *filter p xs* keeps only those elements in xs that satisfy *p*.

 $\begin{array}{ll} filter & :: (a \rightarrow Bool) \rightarrow List \ a \rightarrow List \ a \\ filter \ p \ [] & = [] \\ filter \ p \ (x : xs) \ | \ p \ x = x : filter \ p \ xs \\ | \ otherwise = filter \ p \ xs \ . \end{array}$

TAKE AND DROP

• Recall *take* and *drop*, which we used in the previous exercise.

take :: Nat \rightarrow List $a \rightarrow$ List a take 0 xs = [] take $(1_+ n) [] = []$ $take (1_{+} n) (x : xs) = x : take n xs$. $:: Nat \rightarrow List a \rightarrow List a$ drop drop 0 xs = XS $drop(1_{+} n)[] = []$ $drop(\mathbf{1}_{+} n)(x:xs) = drop n xs$.

• Prove: take n xs + drop n xs = xs, for all n and xs.

TAKEWHILE AND DROPWHILE

• *takeWhile p xs* yields the longest prefix of *xs* such that *p* holds for each element.

takeWhile:: $(a \rightarrow Bool) \rightarrow List \ a \rightarrow List \ a$ takeWhile p []= []takeWhile p (x : xs)| $p \ x = x : takeWhile \ p \ xs$ | otherwise = [].

• *dropWhile p xs* drops the prefix from xs.

 $\begin{array}{ll} dropWhile & :: (a \rightarrow Bool) \rightarrow List \ a \rightarrow List \ a \\ dropWhile \ p \ [] & = [] \\ dropWhile \ p \ (x : xs) \ | \ p \ x = dropWhile \ p \ xs \\ | \ otherwise = x : xs \ . \end{array}$

• Prove: takeWhile p xs ++ dropWhile p xs = xs.

LIST REVERSAL

• reverse [1, 2, 3, 4] = [4, 3, 2, 1].

 $\begin{array}{ll} reverse & :: List \ a \to List \ a \\ reverse \ [] & = \ [] \\ reverse \ (x : xs) & = reverse \ xs + [x] \end{array}.$

ALL PREFIXES AND SUFFIXES

• inits [1, 2, 3] = [[], [1], [1, 2], [1, 2, 3]]inits :: List a → List (List a) inits [] = [[]] inits (x : xs) = [] : map (x :) (inits xs) . • tails [1, 2, 3] = [[1, 2, 3], [2, 3], [3], [3]] tails :: List a → List (List a) tails [] = [[]] tails (x : xs) = (x : xs) : tails xs .

TOTALITY

• Structure of our definitions so far:

 $f[] = \dots$ $f(x : xs) = \dots f xs \dots$

- Both the empty and the non-empty cases are covered, guaranteeing there is a matching clause for all inputs.
- The recursive call is made on a "smaller" argument, guranteeing termination.
- Together they guarantee that every input is mapped to some output. Thus they define *total* functions on lists.

• Some functions discriminate between several base cases. E.g.

 $\begin{array}{ll} fib & :: Nat \rightarrow Nat \\ fib & 0 & = 0 \\ fib & 1 & = 1 \\ fib & (2+n) & = fib & (\mathbf{1}_{+}n) + fib & n \end{array} .$

• Some functions make more sense when it is defined only on non-empty lists:

 $f[x] = \dots$ $f(x : xs) = \dots$

- What about totality?
 - They are in fact functions defined on a different datatype:

data $List^+ a = Singleton a \mid a : List^+ a$.

- We do not want to define *map*, *filter* again for *List*⁺ *a*. Thus we reuse *List a* and pretend that we were talking about *List*⁺ *a*.
- · It's the same with Nat. We embedded Nat into Int.
- Ideally we'd like to have some form of *subtyping*. But that makes the type system more complex.

LEXICOGRAPHIC INDUCTION

- It also occurs often that we perform *lexicographic induction* on multiple arguments: some arguments decrease in size, while others stay the same.
- E.g. the function *merge* merges two sorted lists into one sorted list:

 $\begin{array}{ll} merge & :: List Int \rightarrow List Int \rightarrow List Int \\ merge [] [] & = [] \\ merge [] (y:ys) & = y:ys \\ merge (x:xs) [] & = x:xs \\ merge (x:xs) (y:ys) | x \leq y = x:merge xs (y:ys) \\ | otherwise = y:merge (x:xs) ys . \end{array}$

Another example:

```
\begin{aligned} zip :: \text{List } a \to \text{List } b \to \text{List } (a, b) \\ zip [] [] &= [] \\ zip [] (y : ys) &= [] \\ zip (x : xs) [] &= [] \\ zip (x : xs) (y : ys) &= (x, y) : zip xs ys . \end{aligned}
```

NON-STRUCTURAL INDUCTION

- In most of the programs we've seen so far, the recursive call are made on direct sub-components of the input (e.g. f(x : xs) = ..f xs..). This is called *structural induction*.
 - It is relatively easy for compilers to recognise structural induction and determine that a program terminates.
- In fact, we can be sure that a program terminates if the arguments get "smaller" under some (well-founded) ordering.

MERGESORT

• In the implemenation of mergesort below, for example, the arguments always get smaller in size.

 $\begin{array}{ll} msort & :: List Int \rightarrow List Int \\ msort [] &= [] \\ msort [x] &= [x] \\ msort xs &= merge \ (msort ys) \ (msort zs) \ , \\ \mbox{where } n &= length \ xs \ 'div' \ 2 \\ ys &= take \ n \ xs \\ zs &= drop \ n \ xs \ . \end{array}$

- What if we omit the case for [x]?
- If all cases are covered, and all recursive calls are applied to smaller arguments, the program defines a total function.

• Example of a function, where the argument to the recursive does not reduce in size:

 $\begin{aligned} f & :: Int \to Int \\ f0 & = 0 \\ fn & = fn . \end{aligned}$

• Certainly *f* is not a total function. Do such definitions "mean" something? We will talk about these later.

USER DEFINED INDUCTIVE DATATYPES

INTERNALLY LABELLED BINARY TREES

• This is a possible definition of internally labelled binary trees:

data Tree a = Null | Node a (Tree a) (Tree a) ,

• on which we may inductively define functions:

sumT:: Tree Nat \rightarrow NatsumT Null= 0sumT (Node x t u)= x + sumT t + sumT u .

Exercise: given $(\downarrow) :: Nat \rightarrow Nat \rightarrow Nat$, which yields the smaller one of its arguments, define the following functions

- 1. *minT* :: *Tree Nat* \rightarrow *Nat*, which computes the minimal element in a tree.
- 2. $mapT :: (a \rightarrow b) \rightarrow Tree \ a \rightarrow Tree \ b$, which applies the functional argument to each element in a tree.
- 3. Can you define (\downarrow) inductively on *Nat*? ⁴

⁴In the standard Haskell library, (\downarrow) is called *min*.

INDUCTION PRINCIPLE FOR Tree

- What is the induction principle for *Tree*?
- To prove that a predicate *P* on *Tree* holds for every tree, it is sufficient to show that

INDUCTION PRINCIPLE FOR Tree

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- To prove that a predicate *P* on *Tree* holds for every tree, it is sufficient to show that
 - 1. P Null holds, and;
 - 2. for every x, t, and u, if P t and P u holds, P (Node x t u) holds.

INDUCTION PRINCIPLE FOR Tree

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- To prove that a predicate *P* on *Tree* holds for every tree, it is sufficient to show that
 - 1. P Null holds, and;
 - 2. for every *x*, *t*, and *u*, if *P t* and *P u* holds, *P* (Node *x t u*) holds.
- Exercise: prove that for all n and t, minT (mapT (n+) t) = n + minT t. That is, minT · mapT (n+) = (n+) · minT.

INDUCTION PRINCIPLE FOR OTHER TYPES

- Recall that **data** *Bool* = *False* | *True*. Do we have an induction principle for *Bool*?
- To prove a predicate *P* on *Bool* holds for all booleans, it is sufficient to show that

INDUCTION PRINCIPLE FOR OTHER TYPES

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- To prove a predicate *P* on *Bool* holds for all booleans, it is sufficient to show that
 - 1. P False holds, and
 - 2. *P True* holds.
- Well, of course.

- What about $(A \times B)$? How to prove that a predicate *P* on $(A \times B)$ is always true?
- One may prove some property P_1 on A and some property P_2 on B, which together imply P.
- That does not say much. But the "induction principle" for products allows us to extract, from a proof of *P*, the proofs *P*₁ and *P*₂.

- Every inductively defined datatype comes with its induction principle.
- We will come back to this point later.