

# COMP3141

## Software System Design and Implementation

### Introduction

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Term 2 2019

# What is this course?

Software must be high quality:  
**correct, safe and secure.**

Software must developed  
**cheaply and quickly**



# Safety-uncritical Applications



**Video games:** Some bugs are acceptable, to save developer effort.

# Safety-critical Applications

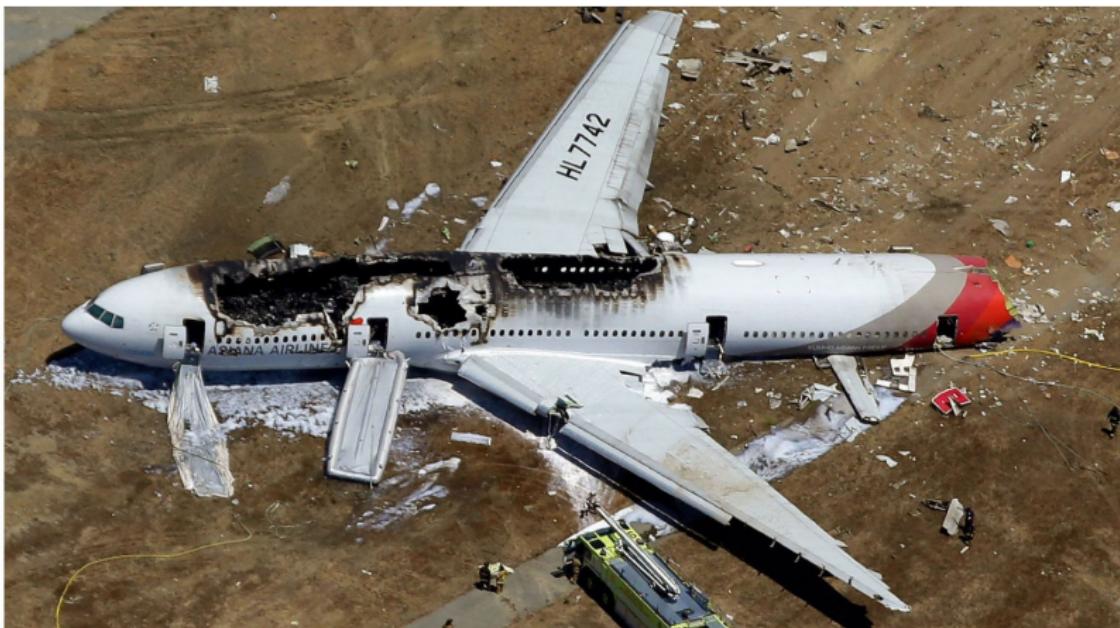
Imagine you...

- Are travelling on a plane
  - Are travelling in a self-driving car
  - Are working on a Mars probe
  - Have invested in a new hedge fund
  - Are running a cryptocurrency exchange
  - Are getting treatment from a radiation therapy machine
  - Intend to launch some nuclear missiles at your enemies
- ... running on software written by **the person next to you.**

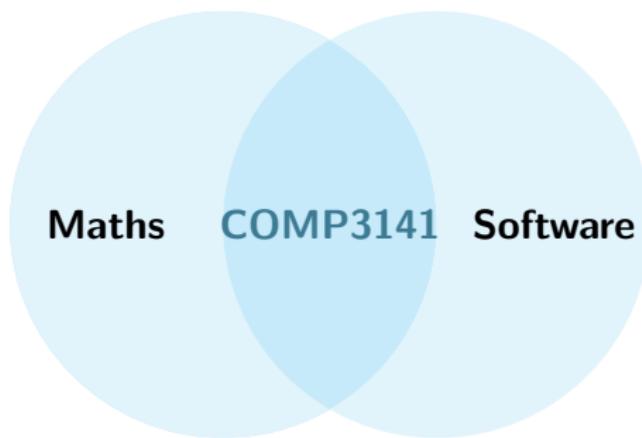
# Safety-critical Applications

## *Airline Blames Bad Software in San Francisco Crash*

The New York Times



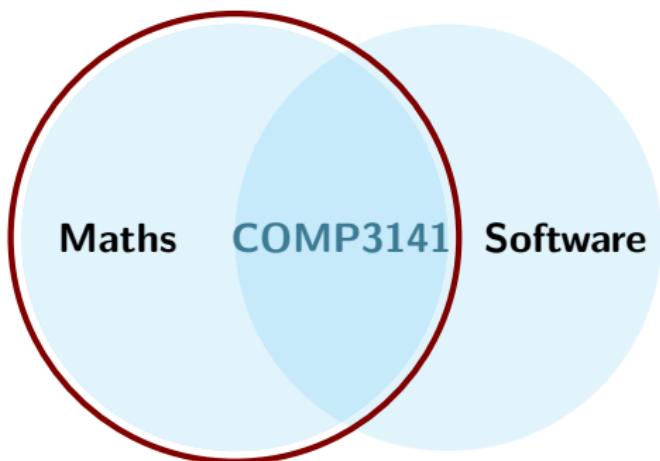
# What is this course?



# What is this course?

## Maths?

- Logic
- Set Theory
- Proofs
- Induction
- Algebra (a bit)
- No calculus ☺

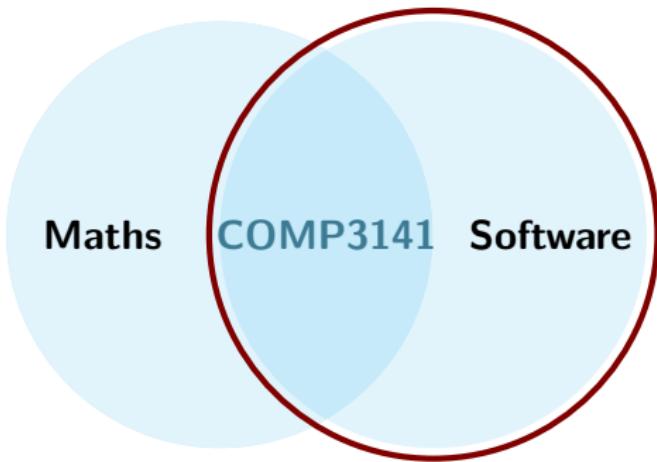


N.B: MATH1081 is neither necessary nor sufficient for COMP3141.

# What is this course?

## Software?

- Programming
- Reasoning
- Design
- Testing
- Types
- Haskell



N.B: Haskell knowledge is not a prerequisite for COMP3141.

# What isn't this course?

This course is **not**:

- a Haskell course
- a formal verification course (for that, see COMP4161)
- an OOP software design course (see COMP2511, COMP1531)
- a programming languages course (see COMP3161).
- a WAM booster cakewalk (hopefully).
- a soul-destroying nightmare (hopefully).

# Assessment

## Warning

For many of you, this course will present a lot of new topics. Even if you are a seasoned programmer, you may have to learn as if from scratch.

- Class Mark (out of 100)
  - **Two** programming assignments, each worth 20 marks.
  - Weekly online quizzes, worth 20 marks.
  - Weekly programming exercises, worth 40 marks.
- Final Exam Mark (out of 100)

$$\text{result} = \frac{2 \cdot \text{class} \cdot \text{exam}}{\text{class} + \text{exam}}$$

# Lectures

- There is one stream of lectures, and a (<sup>hurk!</sup>) web stream.
- I will generally run lectures introducing new material on Tuesday, and Christine will reinforce this material with questions and examples on Wednesday.
- Web stream students **must** watch recordings as they come out.
- Recordings are available through echo 360. I will try my best to make these usable.
- All board-work will be done digitally and made available to you.
- Online quizzes are due one week after the lectures they examine, but **do them early!**

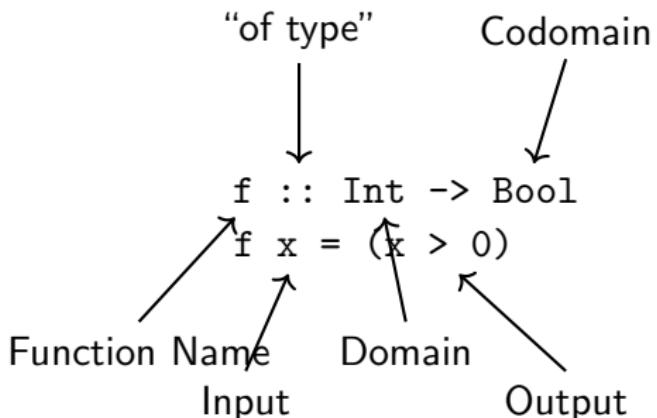
## Books

There are no set textbooks for this course, however there are various books that are useful for learning Haskell listed on the course website.

I can also provide more specialised text recommendations for specific topics.

# Haskell

In this course we use Haskell, because it is the most widespread language with good support for mathematically structured programming.



In mathematics, we would apply a function by writing  $f(x)$ . In Haskell we write `f x`.

**Demo: GHCi, basic functions**

# Currying

- In mathematics, we treat  $\log_{10}(x)$  and  $\log_2(x)$  and  $\ln(x)$  as separate functions.
- In Haskell, we have a single function `logBase` that, given a number  $n$ , produces a function for  $\log_n(x)$ .

```
log10 :: Double -> Double
```

```
log10 = logBase 10
```

```
log2 :: Double -> Double
```

```
log2 = logBase 2
```

```
ln :: Double -> Double
```

```
ln = logBase 2.71828
```

What's the **type** of `logBase`?

# Currying and Partial Application

`logBase :: Double -> (Double -> Double)`

(parentheses optional above)

Function application associates to the **left** in Haskell, so:

`logBase 2 64 ≡ (logBase 2) 64`

Functions of more than one argument are usually written this way in Haskell, but it is possible to use **tuples** instead...

# Tuples

Tuples are another way to take multiple inputs or produce multiple outputs:

```
toCartesian :: (Double, Double) -> (Double, Double)
toCartesian (r, theta) = (x, y)
  where x = r * cos theta
        y = r * sin theta
```

N.B: The order of bindings doesn't matter. Haskell functions have no side effects, they just return a result.

# Higher Order Functions

In addition to returning functions, functions can take other functions as arguments:

```
twice :: (a -> a) -> (a -> a)
twice f a = f (f a)
```

```
double :: Int -> Int
double x = x * 2
```

```
quadruple :: Int -> Int
quadruple = twice double
```

# Lists

Haskell makes extensive use of lists, constructed using square brackets. Each list element must be of the same type.

```
[True, False, True]    ::  [Bool]  
[3, 2, 5+1]           ::  [Int]  
[sin, cos]            ::  [Double -> Double]  
[(3,'a'),(4,'b')]   ::  [(Int, Char)]
```

# Map

A useful function is `map`, which, given a function, applies it to each element of a list:

```
map not [True, False, True] = [False, True, False]
map negate [3, -2, 4]       = [-3, 2, -4]
map (\x -> x + 1) [1, 2, 3] = [2, 3, 4]
```

The last example here uses a *lambda expression* to define a one-use function without giving it a name.

What's the type of `map`?

```
map :: (a -> b) -> [a] -> [b]
```

# Strings

The type `String` in Haskell is just a list of characters:

```
type String = [Char]
```

This is a *type synonym*, like a `typedef` in C.

Thus:

```
"hi!" == ['h', 'i', '!']
```

# Word Frequencies

Let's solve a problem to get some practice:

## Example (First Demo Task)

Given a number  $n$  and a string  $s$ , generate a report (in String form) that lists the  $n$  most common words in the string  $s$ .

We must:

- ➊ Break the input string into words.
- ➋ Convert the words to lowercase.
- ➌ Sort the words.
- ➍ Count adjacent runs of the same word.
- ➎ Sort by size of the run.
- ➏ Take the first  $n$  runs in the sorted list.
- ➐ Generate a report.

# Function Composition

We used *function composition* to combine our functions together.

The mathematical  $(f \circ g)(x)$  is written  $(f . g) x$  in Haskell.

In Haskell, operators like function composition are themselves functions. You can define your own!

```
-- Vector addition
```

```
(.+) :: (Int, Int) -> (Int, Int) -> (Int, Int)
(x1, y1) .+ (x2, y2) = (x1 + x2, y1 + y2)
```

```
(2,3) .+ (1,1) == (3,4)
```

You could even have defined function composition yourself if it didn't already exist:

```
(.) :: (b -> c) -> (a -> b) -> (a -> c)
(f . g) x = f (g x)
```

## Lists

How were all of those list functions we just used implemented?

Lists are **singly-linked** lists in Haskell. The empty list is written as `[]` and a list node is written as `x : xs`. The value `x` is called the **head** and the rest of the list `xs` is called the **tail**. Thus:

```
"hi!" == ['h', 'i', '!'] == 'h':('i':('!':[]))  
                      == 'h' : 'i' : '!' : []
```

When we define recursive functions on lists, we use the last form for pattern matching:

```
map :: (a -> b) -> [a] -> [b]  
map f []      = []  
map f (x:xs) = f x : map f xs
```

# Equational Evaluation

```
map f []      = []
map f (x:xs) = f x : map f xs
```

We can evaluate programs *equationally*:

```
map toUpper "hi!"    ≡ map toUpper ('h':"i!")
                     ≡ toUpper 'h' : map toUpper "i!"
                     ≡ 'H' : map toUpper "i!"
                     ≡ 'H' : map toUpper ('i':"!")
                     ≡ 'H' : toUpper 'i' : map toUpper "!"
                     ≡ 'H' : 'I' : map toUpper "!"
                     ≡ 'H' : 'I' : map toUpper ('!':(""))
                     ≡ 'H' : 'I' : '!' : map toUpper ""
                     ≡ 'H' : 'I' : '!' : map toUpper []
                     ≡ 'H' : 'I' : '!' : []
                     ≡ "HI!"
```

# Higher Order Functions

The rest of this lecture will be spent introducing various list functions that are built into Haskell's standard library by way of [live coding](#).

## Functions to cover:

- ➊ map
- ➋ filter
- ➌ concat
- ➍ sum
- ➎ foldr
- ➏ foldl

In the process, we will introduce **let** and **case** syntax, **guards** and **if**, and the **\$** operator.

# COMP3141

Software System Design and Implementation

## Functional Programming Practice

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Term 2 2019

# Recap: What is this course?

Software must be high quality:  
**correct, safe and secure.**

Software must developed  
**cheaply and quickly**



# Recall: Safety-critical Applications

For safety-critical applications, failure is not an option:

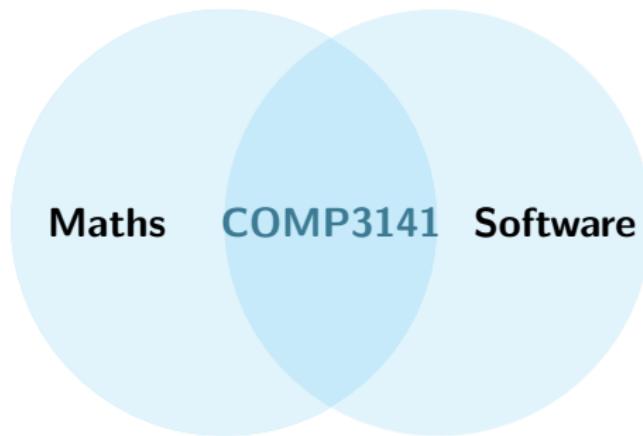
- planes, self-driving cars
- rockets, Mars probe
- drones, nuclear missiles
- banks, hedge funds, cryptocurrency exchanges
- radiation therapy machines, artificial cardiac pacemakers

# Safety-critical Applications



A bug in the code controlling the Therac-25 radiation therapy machine was directly responsible for at least five patient deaths in the 1980s when it administered excessive quantities of beta radiation.

# COMP3141: Functional Programming



# Functional Programming: How does it Help?

- ➊ Close to Maths: more abstract, less error-prone
- ➋ Types: act as doc., the compiler eliminates many errors
- ➌ Property-Based Testing: QuickCheck (in Week 3)
- ➍ Verification: equational reasoning eases proofs (in Week 4)

# COMP3141: Learning Outcomes

- ➊ Identify basic Haskell **type errors** involving concrete types.
- ➋ Work comfortably with **GHCi** on your working machine.
- ➌ Use Haskell **syntax** such as guards, **let**-bindings, **where** blocks, **if** etc.
- ➍ Understand the **precedence of function application** in Haskell, the `(.)` and `(\$)` operators.
- ➎ Write Haskell programs to manipulate **lists** with recursion.
- ➏ Makes use of **higher order functions** like *map* and *fold*.
- ➐ Use  **$\lambda$ -abstraction** to define anonymous functions.
- ➑ Write Haskell programs to compute **basic arithmetic, character, and string manipulation**.
- ➒ Decompose problems using **bottom-up design**.

# Functional Programming: History in Academia

- 1930s** Alonzo Church developed lambda calculus  
(equiv. to Turing Machines)
- 1950s** John McCarthy developed Lisp (LISt Processor, first FP language)
- 1960s** Peter Landin developed ISWIM (If you See What I Mean, first pure FP language)
- 1970s** John Backus developed FP (Functional Programming, higher-order functions, reasoning)
- 1970s** Robin Milner and others developed ML (Meta-Language, first modern FP language, polymorphic types, type inference)
- 1980s** David Turner developed Miranda (lazy, predecessor of Haskell)
- 1987-** An international PL committee developed Haskell (named after the logician Curry Haskell)
  - ... received Turing Awards (similar to Nobel prize in CS).
  - Functional programming is now taught at most CS departments.

# Functional Programming: Influence In Industry

- Facebook's motto was:
  - "Move fast and break things."
  - as they expanded, they understood the importance of bug-free software
  - now Facebook uses functional programming!
- JaneStreet, Facebook, Google, Microsoft, Intel, Apple  
(... and the list goes on)
- Facebook building React and Reason, Apple pivoting to Swift,  
Google developing MapReduce.

# Closer to Maths: Quicksort Example

Let's solve a problem to get some practice:

## Example (Quicksort, recall from Algorithms)

Quicksort is a divide and conquer algorithm.

- ➊ Picks a pivot from the array or list
- ➋ Divides the array or list into two smaller sub-components: the smaller elements and the larger elements.
- ➌ Recursively sorts the sub-components.

- What is the average complexity of Quicksort?
- What is the worst case complexity of Quicksort?
- Imperative programs describe **how** the program works.
- Functional programs describe **what** the program does.

# Quicksort Example (Imperative)

```
algorithm quicksort(A, lo, hi) is
    if lo < hi then
        p := partition(A, lo, hi)
        quicksort(A, lo, p - 1)
        quicksort(A, p + 1, hi)
```

```
algorithm partition(A, lo, hi) is
    pivot := A[hi]
    i := lo
    for j := lo to hi - 1 do
        if A[j] < pivot then
            swap A[i] with A[j]
            i := i + 1
    swap A[i] with A[hi]
    return i
```

# Quick Sort Example (Functional)

```
qsort :: Ord a => [a] -> [a]
qsort [] = []
qsort (x:xs) = qsort smaller ++ [x] ++ qsort larger
    where
        smaller = filter (\ a-> a <= x) xs
        larger = filter (\ b-> b > x) xs
```

Is that it? Does this work?

# Practice Types

In the previous lecture, you learned about the importance of types in functional programming. Let's practice figuring out the types of terms.

- ➊ `True :: Bool`
- ➋ `'a' :: Char`
- ➌ `['a', 'b', 'c'] :: [Char]`
- ➍ `"abc" :: [Char]`
- ➎ `["abc"] :: [[Char]]`
- ➏ `[('f',True), ('e', False)] :: [(Char, Bool)]`

- In Haskell and GHCI using :t.
- Using Haskell documentation and GHCI, answer the questions in this week's quiz (assessed!).

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- ➏ Makes use of **higher order functions** like *map* and *fold*.
- ➐ Use  **$\lambda$ -abstraction** to define anonymous functions.
- ➑ Write Haskell programs to compute **basic arithmetic**, **character**, and **string manipulation**.
- ➒ Decompose problems using **bottom-up design**.

# Recall: Higher Order List Functions

The rest of last lecture was spent introducing various list functions that are built into Haskell's standard library by way of **live coding**.

## Functions covered:

- ① map
- ② filter
- ③ concat
- ④ sum
- ⑤ foldr
- ⑥ foldl

In the process, you saw **guards** and **if**, and the **.** operator.

# Higher Order List Functions

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## Functions covered:

- ① `map`
- ② `filter`
- ③ `concat`
- ④ `sum`
- ⑤ `foldr`
- ⑥ `foldl`

In the process, you saw **guards** and **if**, and the **.** operator.

Let's do that again in Haskell.

# COMP3141: Learning Outcomes

- ➊ Identify basic Haskell **type errors** involving concrete types.
- ➋ Work comfortably with **GHCi** on your working machine.
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# Numbers into Words

Let's solve a problem to get some practice:

## Example (Demo Task)

Given a number  $n$ , such that  $0 \leq n < 1000000$ , generate words (in String form) that describes the number  $n$ .

We must:

- ➊ Convert single-digit numbers into words ( $0 \leq n < 10$ ).
- ➋ Convert double-digit numbers into words ( $0 \leq n < 100$ ).
- ➌ Convert triple-digit numbers into words ( $0 \leq n < 1000$ ).
- ➍ Convert hexa-digit numbers into words ( $0 \leq n < 1000000$ ).

# Single Digit Numbers into Words

$$0 \leq n < 10$$

```
units :: [String]
units =
    ["zero", "one", "two", "three", "four", "five",
     "six", "seven", "eight", "nine", "ten"]

convert1 :: Int -> String

convert1 n = units !! n
```

# Double Digit Numbers into Words

$$0 \leq n < 100$$

```
teens :: [String]
teens =
  ["ten", "eleven", "twelve", "thirteen", "fourteen",
   "fifteen", "sixteen", "seventeen", "eighteen",
   "nineteen"]
```

```
tens :: [String]
tens =
  ["twenty", "thirty", "fourty", "fifty", "sixty",
   "seventy", "eighty", "ninety"]
```

# Double Digit Numbers into Words Continued

$$(0 \leq n < 100)$$

```
digits2 :: Int -> (Int, Int)
digits2 n = (div n 10, mod n 10)
combine2 :: (Int, Int) -> String
combine2 (t, u)
  | t == 0          = convert1 u
  | t == 1          = teens !! u
  | t > 1 && u == 0 = tens !! (t-2)
  | t > 1 && u /= 0 = tens !! (t-2)
                           ++ "-" ++ convert1 u
convert2 :: Int -> String
convert2 = combine2 . digits2
```

# Infix Notation

Instead of

```
digits2 n = (div n 10, mod n 10)
```

for **infix** notation, write:

```
digits2 n = (n `div` 10, n `mod` 10)
```

Note: this is not the same as single quote used for Char ('a').

## Simpler Guards but Order Matters

You could also simplify the guards as follows:

```
combine2 :: (Int, Int) -> String
combine2 (t,u)
| t == 0      = convert1 u
| t == 1      = teens !! u
| u == 0      = tens !! (t-2)
| otherwise   = tens !! (t-2) ++ "—" ++ convert1 u
```

but now the order in which we write the equations is crucial.  
otherwise is a synonym for True.

# Where instead of Function Composition

Instead of implementing convert2 as digit2.combine2, we can implement it directly using the where keyword:

```
convert2 :: Int -> String
convert2 n
| t == 0      = convert1 u
| t == 1      = teens !! u
| u == 0      = tens !! (t-2)
| otherwise   = tens !! (t-2) ++ "—" ++ convert1 u
where (t, u) = (n `div` 10, n `mod` 10)
```

# Triple Digit Numbers into Words

( $0 \leq n < 1000$ )

```
convert3 :: Int -> String

convert3 n
| h == 0      = convert2 n
| t == 0      = convert1 h ++ "hundred"
| otherwise   = convert1 h ++ " hundred and "
                  ++ convert2 t
where (h, t) = (n `div` 100, n `mod` 100)
```

# Hexa Digit Numbers into Words

( $0 \leq n < 1000000$ )

```
convert6 :: Int -> String

convert6 n
| m == 0      = convert3 n
| h == 0      = convert3 m ++ "thousand"
| otherwise   = convert3 m ++ link h ++ convert3 h
where (m, h) = (n `div` 1000, n `mod` 1000)

link :: Int -> String
link h = if (h<100) then " and " else " "

convert :: Int -> String
convert = convert6
```

# COMP3141: Learning Outcomes

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- ➒ Decompose problems using **bottom-up design**.

Induction  
○○○○

Data Types  
○○○○○○○○○○

Type Classes  
○○○○○○○○○○○○

Homework  
○

# COMP3141

## Software System Design and Implementation

### Induction, Data Types and Type Classes

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Term 2 2019

## Recap: Induction

Suppose we want to prove that a property  $P(n)$  holds for **all** natural numbers  $n$ .

Remember that the set of natural numbers  $\mathbb{N}$  can be defined as follows:

### Definition of Natural Numbers

- ① 0 is a natural number.
- ② For any natural number  $n$ ,  $n + 1$  is also a natural number.

## Recap: Induction

Therefore, to show  $P(n)$  for all  $n$ , it suffices to show:

- ①  $P(0)$  (the *base case*), and
- ② assuming  $P(k)$  (the *inductive hypothesis*),  
 $\Rightarrow P(k + 1)$  (the *inductive case*).

### Example

Show that  $f(n) = n^2$  for all  $n \in \mathbb{N}$ , where:

$$f(n) = \begin{cases} 0 & \text{if } n = 0 \\ 2n - 1 + f(n - 1) & \text{if } n > 0 \end{cases}$$

(done on iPad)

# Induction on Lists

Haskell lists can be defined similarly to natural numbers.

## Definition of Haskell Lists

- ①  $[]$  is a list.
- ② For any list  $xs$ ,  $x:xs$  is also a list (for any item  $x$ ).

This means, if we want to prove that a property  $P(ls)$  holds for all lists  $ls$ , it suffices to show:

- ①  $P([])$  (the base case)
- ②  $P(x:xs)$  for all items  $x$ , assuming the inductive hypothesis  $P(xs)$ .

# Induction on Lists: Example

```
sum :: [Int] -> Int
sum []      = 0           -- 1
sum (x:xs)  = x + sum xs -- 2

foldr :: (a -> b -> b) -> b -> [a] -> b
foldr f z []      = z           -- A
foldr f z (x:xs) = x `f` foldr f z xs -- B
```

## Example

Prove for all ls:

$$\text{sum } \text{ls} == \text{foldr } (+) \ 0 \ \text{ls}$$

(done on iPad)

# Custom Data Types

So far, we have seen **type synonyms** using the type keyword. For a graphics library, we might define:

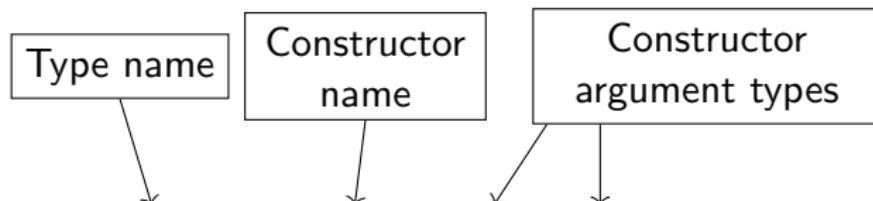
```
type Point    = (Float, Float)
type Vector   = (Float, Float)
type Line     = (Point, Point)
type Colour   = (Int, Int, Int, Int) -- RGBA
```

```
movePoint :: Point -> Vector -> Point
movePoint (x,y) (dx,dy) = (x + dx, y + dy)
```

But these definitions allow Points and Vectors to be used interchangeably, increasing the **likelihood of errors**.

# Product Types

We can define our own compound types using the `data` keyword:



```
data Point = Point Float Float  
           deriving (Show, Eq)
```

```
data Vector = Vector Float Float  
           deriving (Show, Eq)
```

```
movePoint :: Point -> Vector -> Point  
movePoint (Point x y) (Vector dx dy)  
= Point (x + dx) (y + dy)
```

# Records

We could define Colour similarly:

```
data Colour = Colour Int Int Int Int
```

But this has so many parameters, it's hard to tell which is which. Haskell lets us declare these types as *records*, which is identical to the declaration style on the previous slide, but also gives us projection functions and record syntax:

```
data Colour = Colour { redC      :: Int
                      , greenC     :: Int
                      , blueC      :: Int
                      , opacityC   :: Int
                    } deriving (Show, Eq)
```

Here, the code `redC (Colour 255 128 0 255)` gives 255.

# Enumeration Types

Similar to `enums` in C and Java, we can define types to have one of a set of predefined values:

```
data LineStyle = Solid
    | Dashed
    | Dotted
    deriving (Show, Eq)
```

```
data FillStyle = SolidFill | NoFill
    deriving (Show, Eq)
```

Types with more than one constructor are called *sum types*.

# Algebraic Data Types

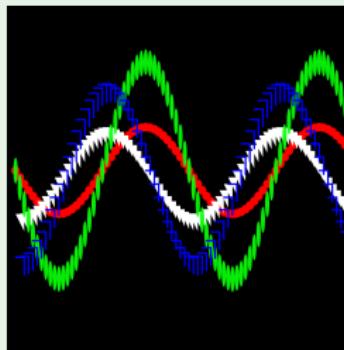
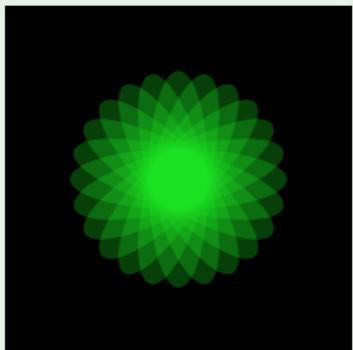
Just as the Point constructor took two Float arguments, constructors for sum types can take parameters too, allowing us to model different kinds of shape:

```
data PictureObject
  = Path      [Point]      Colour LineStyle
  | Circle    Point Float  Colour LineStyle FillStyle
  | Polygon   [Point]      Colour LineStyle FillStyle
  | Ellipse   Point Float  Float  Float
                           Colour LineStyle FillStyle
deriving (Show, Eq)
```

```
type Picture = [PictureObject]
```

# Live Coding: Cool Graphics

## Example (Ellipses and Curves)



# Recursive and Parametric Types

Data types can also be defined with **parameters**, such as the well known Maybe type, defined in the standard library:

```
data Maybe a = Just a | Nothing
```

Types can also be **recursive**. If lists weren't already defined in the standard library, we could define them ourselves:

```
data List a = Nil | Cons a (List a)
```

We can even define natural numbers, where 2 is encoded as Succ(Succ Zero):

```
data Natural = Zero | Succ Natural
```

# Types in Design

## Sage Advice

An old adage due to Yaron Minsky (of Jane Street) is:

*Make illegal states unrepresentable.*

Choose types that *constrain* your implementation as much as possible. Then failure scenarios are eliminated automatically.

## Example (Contact Details)

```
data Contact = C Name (Maybe Address) (Maybe Email)
```

is changed to:

```
data ContactDetails = EmailOnly Email
                     | PostOnly Address
                     | Both Address Email
```

```
data Contact = C Name ContactDetails
```

What failure state is eliminated here? Liam: also talk about other famous screwups

# Partial Functions

Failure to follow Yaron's excellent advice leads to **partial functions**.

## Definition

A **partial function** is a function not defined for all possible inputs.

Examples: head, tail, (!!), division

Partial functions are to be avoided, because they cause your program to crash if undefined cases are encountered.

To eliminate partiality, we must either:

- **enlarge** the codomain, usually with a Maybe type:

```
safeHead :: [a] -> Maybe a -- Q: How is this safer?
```

```
safeHead (x:xs) = Just x
```

```
safeHead []      = Nothing
```

- Or we must **constrain** the domain to be more specific:

```
safeHead' :: NonEmpty a -> a -- Q: How to define?
```

# Type Classes

You have already seen functions such as:

- compare
- (==)
- (+)
- show

that work on **multiple types**, and their corresponding constraints on type variables Ord, Eq, Num and Show.

These constraints are called ***type classes***, and can be thought of as a **set of types** for which certain operations are implemented.

# Show

The Show type class is a set of types that can be converted to strings. It is defined like:

```
class Show a where -- nothing to do with OOP
    show :: a -> String
```

Types are added to the type class as an *instance* like so:

```
instance Show Bool where
    show True = "True"
    show False = "False"
```

We can also define instances that depend on other instances:

```
instance Show a => Show (Maybe a) where
    show (Just x) = "Just " ++ show x
    show Nothing = "Nothing"
```

Fortunately for us, Haskell supports automatically deriving instances for some classes, including Show.

# Read

Type classes can also overload based on the type **returned**, unlike similar features like Java's interfaces:

```
class Read a where
  read :: String -> a
```

Some examples:

- `read "34" :: Int`
- `read "22" :: Char` **Runtime error!**
- `show (read "34") :: String` **Type error!**

# Semigroup

## Semigroups

A *semigroup* is a pair of a set  $S$  and an operation  $\bullet : S \rightarrow S \rightarrow S$  where the operation  $\bullet$  is *associative*.

Associativity is defined as, for all  $a, b, c$ :

$$(a \bullet (b \bullet c)) = ((a \bullet b) \bullet c)$$

Haskell has a type class for semigroups! The associativity law is enforced only by programmer discipline:

```
class Semigroup s where
  (<>>>) :: s -> s -> s
  -- Law: (<>>>) must be associative.
```

What instances can you think of?

# Semigroup

Lets implement additive colour mixing:

```
instance Semigroup Colour where
    Colour r1 g1 b1 a1 <>> Colour r2 g2 b2 a2
        = Colour (mix r1 r2)
                  (mix g1 g2)
                  (mix b1 b2)
                  (mix a1 a2)

    where
        mix x1 x2 = min 255 (x1 + x2)
```

Observe that associativity is satisfied.

# Monoid

## Monoids

A *monoid* is a semigroup  $(S, \bullet)$  equipped with a special *identity element*  $z : S$  such that  $x \bullet z = x$  and  $z \bullet y = y$  for all  $x, y$ .

```
class (Semigroup a) => Monoid a where
    mempty :: a
```

For colours, the identity element is transparent black:

```
instance Monoid Colour where
    mempty = Colour 0 0 0 0
```

For each of the semigroups discussed previously:

- Are they monoids?
- If so, what is the identity element?

Are there any semigroups that are **not** monoids?

## Newtypes

There are multiple possible monoid instances for numeric types like `Integer`:

- The operation `(+)` is associative, with identity element 0
- The operation `(*)` is associative, with identity element 1

Haskell doesn't use any of these, because there can be only **one** instance per type per class in the **entire program** (including all dependencies and libraries used).

A common technique is to define a **separate type** that is represented identically to the original type, but can have its own, different type class instances.

In Haskell, this is done with the `newtype` keyword.

## Newtypes

A newtype declaration is much like a data declaration except that there can be only one constructor and it must take exactly one argument:

```
newtype Score = S Integer
```

```
instance Semigroup Score where
  S x <>> S y = S (x + y)
```

```
instance Monoid Score where
  mempty = S 0
```

Here, `Score` is represented identically to `Integer`, and thus no performance penalty is incurred to convert between them.

In general, newtypes are a great way to prevent mistakes. Use them frequently!

# Ord

Ord is a type class for inequality comparison:

```
class Ord a where
  (<=) :: a -> a -> Bool
```

What laws should instances satisfy?

For all  $x$ ,  $y$ , and  $z$ :

- ① *Reflexivity*:  $x \leq x$ .
- ② *Transitivity*: If  $x \leq y$  and  $y \leq z$  then  $x \leq z$ .
- ③ *Antisymmetry*: If  $x \leq y$  and  $y \leq x$  then  $x = y$ .
- ④ *Totality*: Either  $x \leq y$  or  $y \leq x$

Relations that satisfy these four properties are called *total orders*.  
Without the fourth (totality), they are called *partial orders*.

# Eq

Eq is a type class for equality or equivalence:

```
class Eq a where
  (==) :: a -> a -> Bool
```

What laws should instances satisfy?

For all  $x$ ,  $y$ , and  $z$ :

- ① *Reflexivity*:  $x == x$ .
- ② *Transitivity*: If  $x == y$  and  $y == z$  then  $x == z$ .
- ③ *Symmetry*: If  $x == y$  then  $y == x$ .

Relations that satisfy these are called *equivalence relations*.

Some argue that the Eq class should be only for *equality*, requiring stricter laws like:

If  $x == y$  then  $f\ x == f\ y$  for all functions  $f$

But this is debated.

# COMP3141

Software System Design and Implementation

## Induction, Data Types and Type Classes Practice

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## Recap: Induction

Suppose we want to prove that a property  $P(n)$  holds for **all** natural numbers  $n$ .

Remember that the set of natural numbers  $\mathbb{N}$  can be defined as follows:

### Definition of Natural Numbers

- ① 0 is a natural number.
- ② For any natural number  $n$ ,  $n + 1$  is also a natural number.

Therefore, to show  $P(n)$  for all  $n$ , it suffices to show:

- ①  $P(0)$  (the *base case*), and
- ② assuming  $P(k)$  (the *inductive hypothesis*),  
 $\Rightarrow P(k + 1)$  (the *inductive case*).

# Recap: Induction on Lists

Haskell lists can be defined similarly to natural numbers.

## Definition of Haskell Lists

- ①  $[]$  is a list.
- ② For any list  $xs$ ,  $x:xs$  is also a list (for any item  $x$ ).

This means, if we want to prove that a property  $P(ls)$  holds for all lists  $ls$ , it suffices to show:

- ①  $P([])$  (the base case)
- ②  $P(x:xs)$  for all items  $x$ , assuming the inductive hypothesis  $P(xs)$ .

# Recap: Type Classes

## Semigroups

A *semigroup* is a pair of a set  $S$  and an operation  $\bullet : S \rightarrow S \rightarrow S$  where the operation  $\bullet$  is *associative*.

Associativity is defined as, for all  $a, b, c$ :

$$(a \bullet (b \bullet c)) = ((a \bullet b) \bullet c)$$

# Recap: Type Classes

## Monoids

A *monoid* is a semigroup  $(S, \bullet)$  equipped with a special *identity element*  $z : S$  such that  $x \bullet z = x$  and  $z \bullet y = y$  for all  $x, y$ .

## Example

```
instance Monoid [a] where
    mempty = []
    mappend = (++)
```

# List Monoid Example

```
(++) :: [a] -> [a] -> [a]
(++) []      ys = ys           -- 1
(++) (x:xs) ys = x : xs ++ ys -- 2
```

## Example (Monoid)

Prove for all xs, ys, zs:

$$((xs \text{ ++ } ys) \text{ ++ } zs) = (xs \text{ ++ } (ys \text{ ++ } zs))$$

## Additionally Prove

- ➊ for all xs:

$$[] \text{ ++ } xs == xs$$

- ➋ for all xs:

$$xs \text{ ++ } [] == xs$$

(done on iPad)

# List Reverse Example

```
(++) :: [a] -> [a]
(++) []     ys = ys                                -- 1
(++) (x:xs) ys = x : xs ++ ys                   -- 2
```

```
reverse :: [a] -> [a]
reverse []      = []                               -- A
reverse (x:xs) = reverse xs ++ [x]                -- B
```

## Example

Prove for all ls:

$$\text{reverse}(\text{reverse } \text{ls}) == \text{ls}$$

(done on iPad) stuck!

## List Reverse Example

```
(++) :: [a] -> [a] -> [a]
(++) []     ys = ys                      -- 1
(++) (x:xs) ys = x : xs ++ ys          -- 2
```

```
reverse :: [a] -> [a]
reverse []      = []
reverse (x:xs) = reverse xs ++ [x]       -- A
                                         -- B
```

### Example

To Prove for all ls:

$$\text{reverse}(\text{reverse } \text{ls}) == \text{ls}$$

First Prove for all ys:

$$\text{reverse}(\text{ys} ++ [\text{x}]) = \text{x} : \text{reverse } \text{ys}$$

(done on iPad)

## Recap: Product Type Examples

```
data Point = Point Float Float  
            deriving (Show, Eq)
```

```
data Vector = Vector Float Float  
            deriving (Show, Eq)
```

```
movePoint :: Point -> Vector -> Point  
movePoint (Point x y) (Vector dx dy)  
= Point (x + dx) (y + dy)
```

# Recap: Record Example

```
data Colour = Colour { redC      :: Int
                      , greenC     :: Int
                      , blueC      :: Int
                      , opacityC   :: Int
} deriving (Show, Eq)
```

# Recap: Algebraic Data Types Example

Just as the Point constructor took two Float arguments, constructors for sum types can take parameters too, allowing us to model different kinds of shape:

```
data PictureObject
  = Path      [Point]      Colour LineStyle
  | Circle    Point Float  Colour LineStyle FillStyle
  | Polygon   [Point]      Colour LineStyle FillStyle
  | Ellipse   Point Float  Float  Float
                                Colour LineStyle FillStyle
deriving (Show, Eq)
```

```
type Picture = [PictureObject]
```

# Live Coding: More Cool Graphics

## Example (Fractal Trees)



Property Based Testing  
oooooooooooo

Example  
ooooo

Coverage  
ooo

Lazy Evaluation  
oooo

Homework  
o

# COMP3141

## Software System Design and Implementation

### Property Based Testing; Lazy Evaluation

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Term 2 2019

# Free Properties

Haskell already ensures certain properties automatically with its language design and type system.

- ➊ Memory is accessed where and when it is safe and permitted to be accessed (*memory safety*).
  - ➋ Values of a certain static type will actually have that type at run time.
  - ➌ Programs that are well-typed will not lead to undefined behaviour (*type safety*).
  - ➍ All functions are *pure*: Programs won't have side effects not declared in the type. (*purely functional programming*)
- ⇒ Most of our properties focus on the *logic of our program*.

# Logical Properties

We have already seen a few examples of logical properties.

## Example (Properties)

- ➊ reverse is an *involution*: `reverse (reverse xs) == xs`
- ➋ right identity for `(++)`: `xs ++ [] == xs`
- ➌ transitivity of `(>)`:  $(a > b) \wedge (b > c) \Rightarrow (a > c)$

The set of properties that capture all of our requirements for our program is called the *functional correctness specification* of our software.

This defines what it means for software to be **correct**.

# Proofs

Last week we saw some *proof methods* for Haskell programs. We could **prove** that our implementation meets its functional correctness specification.

Such proofs certainly offer a high degree of assurance, but:

- Proofs must make some assumptions about the environment and the semantics of the software.
- Proof complexity grows with implementation complexity, sometimes drastically.
- If software is **incorrect**, a proof attempt might simply become stuck: we do not always get constructive negative feedback.
- Proofs can be labour and time intensive (\$\$\$), or require highly specialised knowledge (\$\$\$).

# Testing

Compared to proofs:

- Tests typically run the actual program, so requires fewer assumptions about the language semantics or operating environment.
- Test complexity does not grow with implementation complexity, so long as the specification is unchanged.
- Incorrect software when tested leads to immediate, debuggable counterexamples.
- Testing is typically cheaper and faster than proving.
- Tests care about **efficiency** and **computability**, unlike proofs.

We **lose** some assurance, but **gain** some convenience (\$\$\$).

# Property Based Testing

**Key idea:** Generate random input values, and test properties by running them.

## Example (QuickCheck Property)

```
prop_reverseApp xs ys =  
    reverse (xs ++ ys) == reverse ys ++ reverse xs
```

Haskell's *QuickCheck* is the first library ever invented for property-based testing. The concept has since been ported to Erlang, Scheme, Common Lisp, Perl, Python, Ruby, Java, Scala, F#, OCaml, Standard ML, C and C++.

# PBT vs. Unit Testing

- Properties are more compact than unit tests, and describe more cases.  
⇒ Less testing code
- Property-based testing heavily depends on test data generation:
  - Random inputs may not be as informative as hand-crafted inputs  
⇒ use shrinking
  - Random inputs may not cover all necessary corner cases:  
⇒ use a coverage checker
  - Random inputs must be generated for user-defined types:  
⇒ QuickCheck includes functions to build custom generators
- By increasing the number of random inputs, we improve code coverage in PBT.

# Test Data Generation

Data which can be generated randomly is represented by the following type class:

```
class Arbitrary a where
    arbitrary :: Gen a   -- more on this later
    shrink   :: a -> [a]
```

Most of the types we have seen so far implement Arbitrary.

## Shrinking

The shrink function is for when test cases fail. If a given input `x` fails, QuickCheck will try all inputs in `shrink x`; repeating the process until the smallest possible input is found.

# Testable Types

The type of the quickCheck function is:

```
-- more on IO later  
quickCheck :: (Testable a) => a -> IO ()
```

The Testable type class is the class of things that can be converted into properties. This includes:

- Bool values
- QuickCheck's built-in Property type
- Any function from an Arbitrary input to a Testable output:

```
instance (Arbitrary i, Testable o)  
=> Testable (i -> o) ...
```

Thus the type [Int] → [Int] → Bool (as used earlier) is Testable.

## Simple example

Is this function reflexive?

```
divisible :: Integer -> Integer -> Bool  
divisible x y = x `mod` y == 0
```

```
prop_refl :: Integer -> Bool  
prop_refl x = divisible x x
```

- Encode pre-conditions with the ( $\implies$ ) operator:

```
prop_refl :: Integer -> Property  
prop_refl x = x > 0  $\implies$  divisible x x  
(but may generate a lot of spurious cases)
```

- or select different generators with modifier newtypes.

```
prop_refl :: Positive Integer -> Bool  
prop_refl (Positive x) = divisible x x  
(but may require you to define custom generators)
```

# Words and Inverses

## Example (Inverses)

```
words    :: String -> [String]  
unwords :: [String] -> String
```

We might expect unwords to be the inverse of words and vice versa. Let's find out!

**Lessons:** Properties aren't always what you expect!

# Merge Sort

## Example (Merge Sort)

Recall **merge sort**, the sorting algorithm that is reliably  $\mathcal{O}(n \log n)$  time complexity.

- If the list is empty or one element, return that list.
- Otherwise, we:
  - ➊ Split the input list into two sublists.
  - ➋ Recursively sort the two sublists.
  - ➌ Merge the two sorted sublists into one sorted list in linear time.

Applying our bottom up design, let's posit:

```
split :: [a] -> ([a], [a])
merge :: (Ord a) => [a] -> [a] -> [a]
```

# Split

```
split :: [a] -> ([a], [a])
```

What is a good **specification** of split?

- Each element of the input list occurs in one of the two output lists, the same number of times.
- The two output lists consist only of elements from the input list.

Because of its usefulness later, we'll define this in terms of a **permutation** predicate.

# Merge

```
merge :: (Ord a) => [a] -> [a] -> [a]
```

What is a good **specification** of merge?

- Each element of the output list occurs in one of the two input lists, the same number of times.
- The two input lists consist solely of elements from the output list.
- **Important:** If the input lists are sorted, then the output list is sorted.

# Overall

```
mergesort :: (Ord a) => [a] -> [a]
```

What is a good **specification** of mergesort?

- The output list is sorted.
- The output list is a permutation of the input list.

We can prove this as a consequence of the previous specifications which we tested. Do this if time permits.

We can also just write **integration** properties that test the composition of these functions together. Also do this if time permits.

## Redundant Properties

Some properties are technically **redundant** (i.e. implied by other properties in the specification), but there is some value in testing them anyway:

- They may be **more efficient** than full functional correctness tests, consuming less computing resources to test.
- They may be more **fine-grained** to give better test coverage than random inputs for full functional correctness tests.
- They provide a good **sanity check** to the full functional correctness properties.
- Sometimes full functional correctness is **not easily computable** but tests of weaker properties are.

These redundant properties include **unit tests**. We can (and should) combine both approaches!

What are some redundant properties of mergesort?

# Test Quality

How good are your tests?

- Have you checked that every special case works correctly?
- Is all code exercised in the tests?
- Even if all code is exercised, is it exercised in all contexts?

Coverage checkers are useful tools to partially quantify this.

# Types of Coverage

## Branch/Decision Coverage

All conditional branches executed?

## Function Coverage

All functions executed?

## Entry/Exit Coverage

All function calls  
executed?

## Statement/Expression Coverage

All expressions executed?

## Path Coverage

All behaviours executed?  
**very hard!**

# Haskell Program Coverage

Haskell Program Coverage (or `hpc`) is a GHC-bundled tool to measure function, branch and expression coverage.

Let's try it out!

**For Stack:** Build with the `--coverage` flag, execute binary, produce visualisations with `stack hpc report`.

**For Cabal:** Build with the `--enable-coverage` flag, execute binary, produce visualisations with `hpc report`.

# Sum to $n$

```
sumTo :: Integer -> Integer
sumTo 0 = 0
sumTo n = sumTo (n-1) + n
```

This crashes when given a large number. **Why?**

## Sum to $n$ , redux

```
sumTo' :: Integer -> Integer -> Integer
sumTo' a 0 = a
sumTo' a n = sumTo' (a+n) (n-1)

sumTo = sumTo' 0
```

This **still** crashes when given a large number. **Why?**

This is called a **space leak**, and is one of the main drawbacks of Haskell's **lazy evaluation** method.

# Lazy Evaluation

Haskell is **lazily evaluated**, also called **call-by-need**.

This means that expressions are only evaluated when they are **needed** to compute a result for the user.

We can force the previous program to evaluate its accumulator by using a **bang pattern**, or the primitive operation `seq`:

```
sumTo' :: Integer -> Integer -> Integer
```

```
sumTo' !a 0 = a
```

```
sumTo' !a n = sumTo' (a+n) (n-1)
```

```
sumTo' :: Integer -> Integer -> Integer
```

```
sumTo' a 0 = a
```

```
sumTo' a n = let a' = a + n in a' `seq` sumTo' a' (n-1)
```

# Advantages

Lazy Evaluation has many advantages:

- It enables **equational reasoning** even in the presence of partial functions and non-termination.
- It allows functions to be **decomposed** without sacrificing efficiency, for example: `minimum = head . sort` is, depending on sorting algorithm, possibly  $\mathcal{O}(n)$ . John Hughes demonstrates  $\alpha\beta$  pruning from AI as a larger example.<sup>1</sup>
- It allows for **circular programming** and **infinite data structures**, which allow us to express more things as **pure functions**.

## Problem

In **one** pass over a list, replace every element of the list with its maximum.

---

<sup>1</sup>J. Hughes, "Why Functional Programming Matters", Comp. J., 1989

# Infinite Data Structures

Laziness lets us define data structures that extend infinitely. Lists are a common example, but it also applies to trees or any user-defined data type:

```
ones = 1 : ones
```

Many functions such as `take`, `drop`, `head`, `tail`, `filter` and `map` work fine on infinite lists!

```
naturals = 0 : map (1+) naturals
```

--or

```
naturals = map sum (inits ones)
```

How about fibonacci numbers?

```
fibs = 1:1:zipWith (+) fibs (tail fibs)
```

# COMP3141

Software System Design and Implementation

## Property Based Testing Practice

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# Property Based Testing

**Key idea:** Generate random input values, and test properties by running them.

## Example (QuickCheck Property)

```
prop_reverseApp xs ys =  
    reverse (xs ++ ys) == reverse ys ++ reverse xs
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Haskell's *QuickCheck* is the first library ever invented for property-based testing. The concept has since been ported to Erlang, Scheme, Common Lisp, Perl, Python, Ruby, Java, Scala, F#, OCaml, Standard ML, C and C++.

# Mersenne Number Example

## Example (Demo Task)

- The  $n^{th}$  Mersenne number  $M_n = 2^{n-1}$ .
- $M_2, M_3, M_5$  and  $M_7$  are all prime numbers.
- **Conjecture:**  $\forall n.\text{prime}(n) \implies \text{prime}(2^{n-1})$

Let's try using QuickCheck to answer this question.

After 14 guesses and fractions of a second, QuickCheck found a counter-example to this conjecture: 11.

It took humanity about two thousand years to do the same.

# Ransom Note Example

## Example (Demo Task)

Given a magazine (in String form), is it possible to create a ransom message (in String form) from characters in the magazine.

We must:

- ➊ Count the characters in the magazine
- ➋ Count the characters in the message
- ➌ Check whether the magazine contains enough characters to construct the message

In Haskell.

# Recall: Proofs

Proofs:

- Proofs must make some assumptions about the environment and the semantics of the software.
- Proof complexity grows with implementation complexity, sometimes drastically.
- If software is **incorrect**, a proof attempt might simply become stuck: we do not always get constructive negative feedback.
- Proofs can be labour and time intensive (\$\$\$), or require highly specialised knowledge (\$\$\$).

# Testing

Compared to proofs:

- Tests typically run the actual program, so requires fewer assumptions about the language semantics or operating environment.
- Test complexity does not grow with implementation complexity, so long as the specification is unchanged.
- Incorrect software when tested leads to immediate, debuggable counterexamples.
- Testing is typically cheaper and faster than proving.
- Tests care about **efficiency** and **computability**, unlike proofs.

We **lose** some assurance, but **gain** some convenience (\$\$\$).

## Verification versus Validation

*"Testing shows the presence, but not the absence of bugs."*  
– Dijkstra (1969)

Testing is essential but is insufficient for safety-critical applications.

# **COMP3141**

Software System Design and Implementation

# Data Invariants, Abstraction and Refinement

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# Motivation

We've already seen how to **prove** and **test** correctness properties of our programs.

How do we come up with correctness properties in the first place?

# Structure of a Module

A Haskell program will usually be made up of many modules, each of which exports one or more *data types*.

Typically a module for a data type X will also provide a set of functions, called *operations*, on X.

- to construct the data type:  $c :: \dots \rightarrow X$
- to query information from the data type:  $q :: X \rightarrow \dots$
- to update the data type:  $u :: \dots X \rightarrow X$

A lot of software can be designed with this structure.

## Example (Data Types)

A dictionary data type, with empty, insert and lookup.

# Data Invariants

One source of properties is *data invariants*.

## Data Invariants

Data invariants are properties that pertain to a particular data type.

Whenever we use operations on that data type, we want to know that our data invariants are maintained.

## Example

- That a list of words in a dictionary is always in sorted order
- That a binary tree satisfies the search tree properties.
- That a date value will never be invalid (e.g. 31/13/2019).

# Properties for Data Invariants

For a given data type  $X$ , we define a *wellformedness predicate*

$$\text{wf} :: X \rightarrow \text{Bool}$$

For a given value  $x :: X$ ,  $\text{wf } x$  returns true iff our data invariants hold for the value  $x$ .

## Properties

For each operation, if all input values of type  $X$  satisfy  $\text{wf}$ , all output values will satisfy  $\text{wf}$ .

In other words, for each constructor operation  $c :: \dots \rightarrow X$ , we must show  $\text{wf}(c \dots)$ , and for each update operation  $u :: X \rightarrow X$  we must show  $\text{wf } x \implies \text{wf}(u x)$

**Demo:** Dictionary example, sorted order.

# Stopping External Tampering

Even with our sorted dictionary example, there's nothing to stop a malicious or clueless programmer from going in and mucking up our data invariants.

## Example

The malicious programmer could just add a word directly to the dictionary, unsorted, bypassing our carefully written `insert` function.

We want to prevent this sort of thing from happening.

# Abstract Data Types

An *abstract* data type (ADT) is a data type where the implementation details of the type and its associated operations are hidden.

```
newtype Dict
type Word = String
type Definition = String
emptyDict :: Dict
insertWord :: Word -> Definition -> Dict -> Dict
lookup     :: Word -> Dict -> Maybe Definition
```

If we don't have access to the implementation of Dict, then we can only access it via the provided operations, which we know preserve our data invariants. Thus, our data invariants cannot be violated if this module is correct.

**Demo:** In Haskell, we make ADTs with module headers.

# Abstract? Data Types

In general, *abstraction* is the process of **eliminating detail**.

The inverse of abstraction is called *refinement*.

Abstract data types like the dictionary above are **abstract** in the sense that their implementation details are hidden, and we no longer have to reason about them on the level of implementation.

# Validation

Suppose we had a sendEmail function

```
sendEmail :: String -- email address  
           -> String -- message  
           -> IO () -- action (more in 2 wks)
```

It is possible to mix the two String arguments, and even if we get the order right, it's possible that the given email address is not valid.

## Question

Suppose that we wanted to make it impossible to call sendEmail without first checking that the email address was valid.

How would we accomplish this?

# Validation ADTs

We could define a tiny ADT for validated email addresses, where the data invariant is that the contained email address is valid:

```
module EmailADT(Email, checkEmail, sendEmail)
  newtype Email = Email String

  checkEmail :: String -> Maybe Email
  checkEmail str | '@' `elem` str = Just (Email str)
                 | otherwise      = Nothing
```

Then, change the type of sendEmail:

```
sendEmail :: Email -> String -> IO()
```

The only way (outside of the EmailADT module) to create a value of type `Email` is to use `checkEmail`.

`checkEmail` is an example of what we call a *smart constructor*: a constructor that enforces data invariants.

## Reasoning about ADTs

Consider the following, more traditional example of an ADT interface, the unbounded queue:

## data Queue

```
emptyQueue :: Queue
enqueue   :: Int -> Queue -> Queue
front     :: Queue -> Int      -- partial
dequeue   :: Queue -> Queue   -- partial
size      :: Queue -> Int
```

We could try to come up with properties that relate these functions to each other without reference to their implementation, such as:

dequeue (enqueue x emptyQueue) == emptyQueue

However these do not capture functional correctness (usually).

## Models for ADTs

We could imagine a simple implementation for queues, just in terms of lists:

```
emptyQueueL = []
enqueueL a   = (++ [a])
frontL       = head
dequeueL    = tail
sizeL        = length
```

But this implementation is  $\mathcal{O}(n)$  to enqueue! Unacceptable!

### However!

This is a dead simple implementation, and trivial to see that it is correct. If we make a better queue implementation, it should always give the same results as this simple one.

Therefore: This implementation serves as a **functional correctness specification** for our Queue type!

## Refinement Relations

The typical approach to connect our model queue to our Queue type is to define a relation, called a *refinement relation*, that relates a Queue to a list and tells us if the two structures represent the same queue conceptually:

```
rel :: Queue -> [Int] -> Bool
```

Then, we show that the refinement relation holds initially:

```
prop_empty_r = rel emptyQueue emptyQueueL
```

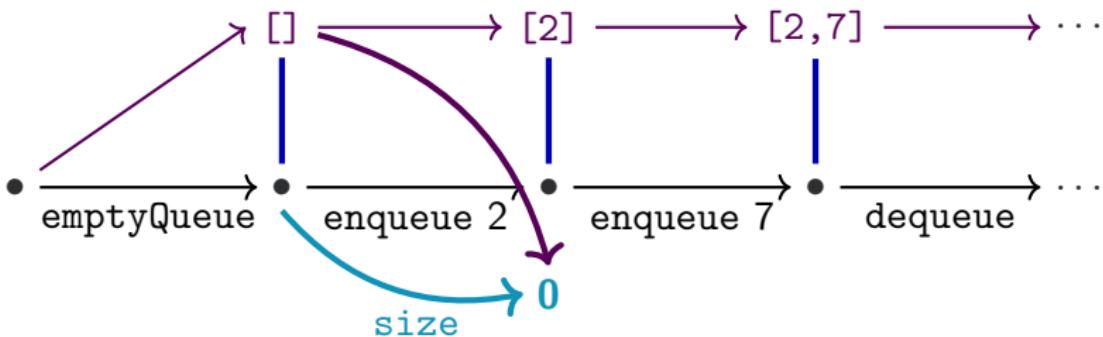
That any query functions for our two types produce equal results for related inputs, such as for size:

```
prop_size_r fq lq = rel fq lq ==> size fq == sizeL lq
```

And that each of the queue operations preserves our refinement relation, for example for enqueue:

```
prop_enq_ref fq lq x =  
  rel fq lq ==> ref (enqueue x fq) (enqueueL x lq)
```

## In Pictures



```
prop_enq_ref fq lq x =  
rel fq lq ==> ref (enqueue x fq) (enqueueL x lq)  
prop_empty_r = rel emptyQueue emptyQueueL  
prop_size_r fq lq = rel fq lq ==> size fq == sizeL lq
```

**Whenever we use a Queue, we can reason as if it were a list!**

## Abstraction Functions

These refinement relations are very difficult to use with QuickCheck because the `rel fq lq` preconditions are very hard to satisfy with randomly generated inputs.

For this example, it's a lot easier if we define an abstraction function that computes the corresponding **abstract** list from the **concrete** Queue.

$$\text{toAbstract} :: \text{Queue} \rightarrow [\text{Int}]$$

Conceptually, our refinement relation is then just:

$$\backslash \text{fq } \text{lq} \rightarrow \text{absfun fq} == \text{lq}$$

However, we can re-express our properties in a much more QC-friendly format (**Demo**)

# Fast Queues

Let's use test-driven development! We'll implement a fast Queue with amortised  $\mathcal{O}(1)$  operations.

```
data Queue = Q [Int] -- front of the queue
              Int    -- size of the front
              [Int] -- rear of the queue
              Int    -- size of the rear
```

We store the rear part of the queue in **reverse order**, to make enqueueing easier.

Thus, converting from our Queue to an abstract list requires us to reverse the rear:

```
toAbstract :: Queue -> [Int]
toAbstract (Q f sf r sr) = f ++ reverse r
```

# Data Refinement

These kinds of properties establish what is known as a *data refinement* from the *abstract*, slow, list model to the fast, *concrete* Queue implementation.

## Refinement and Specifications

In general, all *functional correctness specifications* can be expressed as:

- ① all data invariants are maintained, and
- ② the implementation is a refinement of an abstract correctness model.

There is a limit to the amount of abstraction we can do before they become useless for testing (but not necessarily for proving).

## Warning

While abstraction can simplify proofs, abstraction does not reduce the fundamental complexity of verification, which is provably hard.

## Data Invariants for Queue

In addition to the already-stated refinement properties, we also have some data invariants to maintain for a value  $Q \ f \ sf \ r \ sr$ :

- ①  $\text{length } f == sf$
- ②  $\text{length } r == sr$
- ③ **important:**  $sf \geq sr$  — the front of the queue cannot be shorter than the rear.

We will ensure our `Arbitrary` instance only ever generates values that meet these invariants.

Thus, our `wellformed` predicate is used merely to enforce these data invariants on the outputs of our operations:

```
prop_wf_empty = wellformed (emptyQueue)
prop_wf_enq q = wellformed (enqueue x q)
prop_wf_deq q = size q > 0 ==> wellformed (dequeue q)
```

# Implementing the Queue

We will generally implement by:

- Dequeue from the front.
- Enqueue to the rear.
- If necessary, re-establish the third data invariant by taking the rear, reversing it, and appending it to the front.

This step is slow ( $\mathcal{O}(n)$ ), but only happens every  $n$  operations or so, giving an average case amortised complexity of  $\mathcal{O}(1)$  time.

# Amortised Cost

`enqueue x (Q f sf r sr) = inv3 (Q f sf (x:r) (sr + 1))`

When we enqueue each of [1..7] to the emptyQueue in turn:

Q	[]	0	[]	0
$\rightarrow$	[1]	1	[]	0 (*)
$\rightarrow$	[1]	1	[2]	1
$\rightarrow$	[1, 2, 3]	3	[]	0 (*)
$\rightarrow$	[1, 2, 3]	3	[4]	1
$\rightarrow$	[1, 2, 3]	3	[5, 4]	2
$\rightarrow$	[1, 2, 3]	3	[6, 5, 4]	3
$\rightarrow$	[1, 2, 3, 4, 5, 6, 7]	7	[]	0 (*)

Observe that the slow invariant-reestablishing step (\*) happens after 1 step, then 2, then 4...

Extended out, this averages out to  $\mathcal{O}(1)$ .

## Another Example

Consider this ADT interface for a bag of numbers:

```
data Bag
emptyBag    :: Bag
addToBag    :: Int -> Bag -> Bag
averageBag   :: Bag -> Maybe Int
```

Our conceptual abstract model is just a list of numbers:

```
emptyBagA = []
```

```
addToBagA x xs = x:xs
```

```
averageBagA [] = Nothing
```

```
averageBagA xs = Just (sum xs `div` length xs)
```

But do we need to keep track of all that information in our implementation? **No!**

# Concrete Implementation

Our concrete version will just maintain two integers, the total and the count:

```
data Bag = B { total :: Int , count :: Int }
emptyBag :: Bag
emptyBag = B 0 0
```

```
addToBag :: Int -> Bag -> Bag
addToBag x (B t c) = B (x + t) (c + 1)
```

```
averageBag :: Bag -> Maybe Int
averageBag (B _ 0) = Nothing
averageBag (B t c) = Just (t `div` c)
```

# Refinement Functions

Normally, writing an abstraction function (as we did for Queue) is a good way to express our refinement relation in a QC-friendly way. In this case, however, it's hard to write such a function:

```
toAbstract :: Bag -> [Int]
toAbstract (B t c) = ??????
```

Instead, we will go in the other direction, giving us a *refinement function*:

```
toConc :: [Int] -> Bag
toConc xs = B (sum xs) (length xs)
```

# Properties with Refinement Functions

Refinement functions produce properties much like abstraction functions, only with the abstract and concrete layers swapped:

```
prop_ref_empty =
```

```
  toConc emptyBagA == emptyBag
```

```
prop_ref_add x ab =
```

```
  toConc (addToBagA x ab) == addToBag x (toConc ab)
```

```
prop_ref_avg ab =
```

```
  averageBagA ab == averageBag (toConc ab)
```

# COMP3141

Software System Design and Implementation

## Data Invariants, Abstraction and Refinement Practice

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# Recall: Structure, Data Invariants and Refinement

- **Structure:** a module for a data type X will typically provide operations to construct, query, and update X.
- **Data Invariants:** are properties that pertain to a particular data type that the data type's operations should maintain.
- **Data Refinement:** for every behaviour exhibited by the concrete implementation, there is a similar behaviour in the abstract model.

## Refinement and Specifications

In general, all **functional correctness specifications** can be expressed as:

- ① all data invariants are maintained, and
- ② the implementation is a refinement of an abstract correctness model.

## Editor Example

Consider this ADT interface for a text editor:

```
data Editor
einit :: String -> Editor
stringOf :: Editor -> String
moveLeft :: Editor -> Editor
moveRight :: Editor -> Editor
insertChar :: Char -> Editor -> Editor
deleteChar :: Editor -> Editor
```

# Data Invariant Properties

prop_einit_ok	s = wellformed (einitA s)
prop_moveLeft_ok	a = wellformed (moveLeftA a)
prop_moveRight_ok	a = wellformed (moveRightA a)
prop_moveInsert_ok	x a = wellformed (insertCharA x a)
prop_moveDelete_ok	a = wellformed (deleteCharA a)

## Editor Example: Abstract Model

Our conceptual abstract model is a string and a cursor position:

```
einitA s = A s 0
stringOfA (A s _) = s
moveLeftA (A t c) = A t (max 0 (c-1))
moveRightA (A t c) = A t (min (length t) (c+1))
insertCharA x (A t c) = let (t1, t2) = splitAt c t
                           in A (t1 ++ [x] ++ t2) (c+1)
deleteCharA (A t c) = let (t1, t2) = splitAt c t
                           in A (t1 ++ drop 1 t2) c
```

But do we need to keep track of all that information in our implementation? **No!**

## Concrete Implementation

Our concrete version will just maintain two strings, the left part (in reverse) and the right part of the cursor:

```
einit s = C [] s
stringOf (C ls rs) = reverse ls ++ rs
moveLeft (C (l:ls) rs) = C ls (l:rs)
moveLeft c = c
moveRight (C ls (r:rs)) = C (r:ls) rs
moveRight c = c
insertChar x (C ls rs) = C (x:ls) rs
deleteChar (C ls (_:rs)) = C ls rs
deleteChar c = c
```

# Refinement Functions

Abstraction function to express our refinement relation in a QC-friendly way: such a function:

```
toAbstract :: Concrete -> Abstract  
toAbstract (C ls rs) = A (reverse ls ++ rs) (length ls)
```

# Properties with Abstraction Functions

```
prop_init_r s =
  toAbstract (einit s) == einitA s
prop_stringOf_r c =
  stringOf c == stringOfA (toAbstract c)
prop_moveLeft_r c =
  toAbstract (moveLeft c) == moveLeftA (toAbstract c)
prop_moveRight_r c =
  toAbstract (moveRight c) == moveRightA (toAbstract c)
prop_insChar_r x c =
  toAbstract (insertChar x c)
  == insertCharA x (toAbstract c)
prop_delChar_r c =
  toAbstract (deleteChar c) == deleteCharA (toAbstract c)
```

Higher Kinds

○○○

Functors

○○○○○

Applicative Functors

○○○○○○○○○

Monads

○○○○○

# COMP3141

## Software System Design and Implementation

### Functors, Applicatives, and Monads

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# Motivation

We'll be looking at three **very common** abstractions:

- used in functional programming and,
- increasingly, in imperative programming as well.

Unlike many other languages, these abstractions are reified into bona fide type classes in Haskell, where they are often left as mere "design patterns" in other programming languages.

# Types and Values

First, some preliminaries. Haskell is actually comprised of **two languages**.

- The *value-level* language, consisting of expressions such as if, let, 3 etc.
- The *type-level* language, consisting of types Int, Bool, synonyms like String, and type *constructors* like Maybe, (->), [ ] etc.

This type level language itself has a type system!

# Kinds

Just as terms in the value level language are given types, terms in the type level language are given *kinds*.

The most basic kind is written as \*.

- Types such as Int and Bool have kind \*.
- Seeing as Maybe is parameterised by one argument, Maybe has kind  $* \rightarrow *$ : given a type (e.g. Int), it will return a type (Maybe Int).

# Lists

Suppose we have a function:

```
toString :: Int -> String
```

And we also have a function to give us some numbers:

```
getNumbers :: Seed -> [Int]
```

How can I compose `toString` with `getNumbers` to get a function `f` of type `Seed -> [String]`?

**Answer:** we use `map`:

```
f = map toString . getNumbers
```

# Maybe

Suppose we have a function:

```
toString :: Int -> String
```

And we also have a function that may give us a number:

```
tryNumber :: Seed -> Maybe Int
```

How can I compose `toString` with `tryNumber` to get a function `f` of type `Seed -> Maybe String`?

We want a `map` function **but for the Maybe type**:

```
f = maybeMap toString . tryNumber
```

Let's implement it.

# QuickCheck Generators

Recall the `Arbitrary` class has a function:

```
arbitrary :: Gen a
```

The type `Gen` is an **abstract type** for QuickCheck generators.

Suppose we have a function:

```
toString :: Int -> String
```

And we want a generator for `String` (i.e. `Gen String`) that is the result of applying `toString` to arbitrary `Ints`.

It is reasonable to expect that perhaps there is a `genMap` function:

```
genMap :: (a -> b) -> Gen a -> Gen b
```

# Functor

All of these functions are in the interface of a single type class, called **Functor**.

```
class Functor f where
    fmap :: (a -> b) -> f a -> f b
```

Unlike previous type classes we've seen like `Ord` and `Semigroup`, `Functor` is over types of kind `* -> *`.

Instances for:

- Lists
- Maybe
- Gen
- Tuples (how?)
- Functions (how?)

Demonstrate in live-coding

# Functor Laws

The functor type class must obey two laws:

## Functor Laws

- ①  $\text{fmap id} == \text{id}$
- ②  $\text{fmap f . fmap g} == \text{fmap (f . g)}$

In Haskell's type system it's impossible to make a total `fmap` function that satisfies the first law but violates the second.

This is due to *parametricity*, a property we will return to in Week 8 or 9

## Binary Functions

Suppose we want to look up a student's zID and program code using these functions:

```
lookupID :: Name -> Maybe ZID
```

```
lookupProgram :: Name -> Maybe Program
```

And we had a function:

```
makeRecord :: ZID -> Program -> StudentRecord
```

How can we combine these functions to get a function of type Name -> Maybe StudentRecord?

```
lookupRecord :: Name -> Maybe StudentRecord
```

```
lookupRecord n = let zid      = lookupID n
                  program  = lookupProgram n
                  in ?
```

# Binary Map?

We could imagine a binary version of the `maybeMap` function:

```
maybeMap2 :: (a -> b -> c)
            -> Maybe a -> Maybe b -> Maybe c
```

But then, we might need a trinary version.

```
maybeMap3 :: (a -> b -> c -> d)
            -> Maybe a -> Maybe b -> Maybe c -> Maybe d
```

Or even a 4-ary version, 5-ary, 6-ary...

this would quickly become impractical!

# Using Functor

Using `fmap` gets us part of the way there:

```
lookupRecord' :: Name -> Maybe (Program -> StudentRecord)
lookupRecord' n = let zid      = lookupID n
                  program = lookupProgram n
                  in fmap makeRecord zid
                     -- what about program?
```

But, now we have a function inside a `Maybe`.

We need a function to take:

- A `Maybe`-wrapped fn `Maybe (Program -> StudentRecord)`
- A `Maybe`-wrapped argument `Maybe Program`

And apply the function to the argument, giving us a result of type `Maybe StudentRecord?`

# Applicative

This is encapsulated by a subclass of Functor called Applicative:

```
class Functor f => Applicative f where
    pure :: a -> f a
    (<*>) :: f (a -> b) -> f a -> f b
```

Maybe is an instance, so we can use this for lookupRecord:

```
lookupRecord :: Name -> Maybe StudentRecord
lookupRecord n = let zid      = lookupID n
                  program = lookupProgram n
                  in fmap makeRecord zid <*> program
-- or pure makeRecord <*> zid <*> program
```

# Using Applicative

In general, we can take a regular function application:

f a b c d

And apply that function to Maybe (or other Applicative) arguments using this pattern (where `<*>` is left-associative):

pure f `<*>` ma `<*>` mb `<*>` mc `<*>` md

## Relationship to Functor

All law-abiding instances of Applicative are also instances of Functor, by defining:

```
fmap f x = pure f <*> x
```

Sometimes this is written as an infix operator, `<$>`, which allows us to write:

```
pure f <*> ma <*> mb <*> mc <*> md
```

as:

```
f <$> ma <*> mb <*> mc <*> md
```

**Proof exercise:** From the applicative laws (next slide), prove that this implementation of `fmap` obeys the functor laws.

# Applicative laws

-- *Identity*

```
pure id <*> v = v
```

-- *Homomorphism*

```
pure f <*> pure x = pure (f x)
```

-- *Interchange*

```
u <*> pure y = pure ($ y) <*> u
```

-- *Composition*

```
pure (.) <*> u <*> v <*> w = u <*> (v <*> w)
```

These laws are a bit complex, and we certainly don't expect you to memorise them, but pay attention to them when defining instances!

# Applicative Lists

There are **two** ways to implement Applicative for lists:

`(<*>) :: [a -> b] -> [a] -> [b]`

- ➊ Apply each of the given functions to each of the given arguments, concatenating all the results.
- ➋ Apply each function in the list of functions to the corresponding value in the list of arguments.

**Question:** How do we implement pure?

The second one is put behind a newtype (`ZipList`) in the Haskell standard library.

## Other instances

- QuickCheck generators: Gen  
Recall from Wednesday Week 4:

```
data Concrete = C [Char] [Char]  
    deriving (Show, Eq)
```

```
instance Arbitrary Concrete where  
    arbitrary = C <$> arbitrary <*> arbitrary
```

- Functions: ((->) x)
- Tuples: ((,) x) We can't implement pure!

# On to Monads

- Functors are types for containers where we can map pure functions on their contents.
- Applicative Functors are types where we can combine  $n$  containers with a  $n$ -ary function.

The last and most commonly-used higher-kinded abstraction in Haskell programming is the Monad.

## Monads

Monads are types  $m$  where we can *sequentially compose* functions of the form  $a \rightarrow m b$

# Monads

```
class Applicative m => Monad m where
  (=>=) :: m a -> (a -> m b) -> m b
```

Sometimes in old documentation the function `return` is included here, but it is just an alias for `pure`. It has nothing to do with `return` as in C/Java/Python etc.

Consider for:

- Maybe
- Lists
- ( $x \rightarrow$ )
- Gen

# Monad Laws

We can define a composition operator with ( $>>=$ ):

$$(<=<) :: (b \rightarrow m\ c) \rightarrow (a \rightarrow m\ b) \rightarrow (a \rightarrow m\ c)$$
$$(f <=< g)\ x = g\ x >>= f$$

## Monad Laws

$f <=< (g <=< x) == (f <=< g) <=< x$	-- associativity
$\text{pure} <=< f == f$	-- left identity
$f <=< \text{pure} == f$	-- right identity

These are similar to the monoid laws, generalised for multiple types inside the monad. This sort of structure is called a *category* in mathematics.

## Relationship to Applicative

All Monad instances give rise to an Applicative instance, because we can define `<*>` in terms of `>>=`.

```
mf <*> mx = mf >>= \f -> mx >>= \x -> pure (f x)
```

This implementation is already provided for Monads as the `ap` function in `Control.Monad`.

# Examples

## Example (Dice Rolls)

Roll two 6-sided dice, if the difference is  $< 2$ , reroll the second die. Final score is the difference of the two die. What score is most common?

## Example (Partial Functions)

We have a list of student names in a database of type `[(ZID, Name)]`. Given a list of zID's, return a `Maybe [Name]`, where `Nothing` indicates that a zID could not be found.

## Example (Arbitrary Instances)

Define a `Tree` type and a generator for search trees:

```
searchTrees :: Int -> Int -> Generator Tree
```

## Do notation

We've seen how working with monads can be a bit unpleasant.  
Haskell has some notation to increase niceness:

`do x <- y  
 z`      **becomes**      `y >>= \x -> do z`

`do x  
 y`      **becomes**      `x >>= \_ -> do y`

Let's translate our examples!

# COMP3141

Software System Design and Implementation

## Functors, Applicatives, and Monads Practice

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# Recall: Functors, Applicatives, Monads

- Functors are types for containers where we can map pure functions on their contents.
- Applicative Functors are types where we can combine  $n$  containers with an  $n$ -ary function.
- Monads are types  $m$  where we can *sequentially compose* functions of the form  $a \rightarrow m b$ .

# Recall: Functors

```
class Functor f where
  fmap :: (a -> b) -> f a -> f b
```

The functor type class must obey two laws:

## Functor Laws

- ➊  $fmap\ id == id$
- ➋  $fmap\ f\ .\ fmap\ g == fmap\ (f\ .\ g)$

# Recall: Applicatives

```
class Functor f => Applicative f where
    pure :: a -> f a
    ( <*> ) :: f (a -> b) -> f a -> f b
```

The functor type class must obey four additional laws:

## Applicative Laws

- ➊  $\text{pure id } \langle * \rangle v = v$
- ➋  $\text{pure f } \langle * \rangle \text{pure x} = \text{pure (f x)}$
- ➌  $u \langle * \rangle \text{pure y} = \text{pure (\$ y)} \langle * \rangle u$
- ➍  $\text{pure (.) } \langle * \rangle u \langle * \rangle v \langle * \rangle w = u \langle * \rangle (v \langle * \rangle w)$

# Alternative Applicative

It is possible to express Applicative equivalently as:

```
class Functor f => App f where
    pure :: a -> f a
    tuple :: f a -> f b -> f (a,b)
```

## Example (Alternative Applicative)

- ① Using , fmap and pure, let's implement <\*>.
- ② And, using <\*>, fmap and pure, let's implement tuple.

done in Haskell.

**Proof exercise:** Prove that tuple obeys the applicative laws.

## Recall: Monads

```
class Applicative m => Monad m where
  (>>=) :: m a -> (a -> m b) -> m b
```

We can define a composition operator with ( $>>=$ ):

```
(<=<) :: (b -> m c) -> (a -> m b) -> (a -> m c)
(f <=< g) x = g x >>= f
```

The monad type class must obey three additional laws:

### Monad Laws

- ➊  $f <=< (g <=< x) == (f <=< g) <=< x$  (associativity)
- ➋  $\text{pure} <=< f == f$  (left identity)
- ➌  $f <=< \text{pure} == f$  (right identity)

# Alternative Monad

It is possible to express Monad equivalently as:

```
class Applicative m => Mon m where
    join :: m (m a) -> m a
```

## Example (Alternative Monad)

- ① Using join and fmap, let's implement `>>=`.
- ② And, using `>>=` let's implement join.

done in Haskell.

# Tree Example

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

## Example (Tree Example)

Show that Tree is an Applicative instance.  
done in Haskell.

Note that Tree is not a Monad instance.

# Formulas Example

```
data Formula v = Var v
                | Plus (Formula v) (Formula v)
                | Times (Formula v) (Formula v)
                | Constant Int
                deriving (Eq, Show)
```

## Example (Formulas Example)

Show that `Formula` is a Monad instance.  
done in Haskell.

Effects  
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State Monad  
○○

IO Monad  
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QuickChecking Monads  
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# COMP3141

## Software System Design and Implementation

### Effects and State

Liam O'Connor  
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Term 2 2019

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
○

# Effects

## Effects

*Effects* are observable phenomena from the execution of a program.

### Example (Memory effects)

```
int *p = ...  
... // read and write  
*p = *p + 1;
```

### Example (IO)

```
// console IO  
c = getchar();  
printf("%d", 32);
```

### Example (Non-termination)

```
// infinite loop  
while (1) {};
```

### Example (Control flow)

```
// exception effect  
throw new Exception();
```

Effects  
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State Monad  
○○

IO Monad  
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QuickChecking Monads  
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# Internal vs. External Effects

## External Observability

An *external* effect is an effect that is **observable** outside the function.

*Internal* effects are not observable from outside.

## Example (External effects)

Console, file and network I/O; termination and non-termination; non-local control flow; etc.

Are memory effects *external* or *internal*?

**Answer:** Depends on the scope of the memory being accessed.  
Global variable accesses are *external*.

# Purity

A function with no external effects is called a *pure* function.

## Pure functions

A *pure function* is the mathematical notion of a function. That is, a function of type  $a \rightarrow b$  is *fully* specified by a mapping from all elements of the domain type  $a$  to the codomain type  $b$ .

Consequences:

- Two invocations with the same arguments result in the same value.
- No observable trace is left beyond the result of the function.
- No implicit notion of time or order of execution.

**Question:** Are Haskell functions *pure*?

# Haskell Functions

Haskell functions are technically **not** pure.

- They can loop infinitely.
- They can throw exceptions (**partial functions**).
- They can force evaluation of unevaluated expressions.

## Caveat

Purity only applies to a particular level of abstraction. Even ignoring the above, assembly instructions produced by GHC aren't really pure.

Despite the impurity of Haskell functions, we can often reason as though they are pure. Hence we call Haskell a **purely functional** language.

# The Danger of Implicit Side Effects

- They introduce (often subtle) requirements on the evaluation order.
- They are not visible from the type signature of the function.
- They introduce **non-local** dependencies which is bad for software design, increasing ***coupling***.
- They interfere badly with strong typing, for example mutable arrays in Java, or reference types in ML.

We can't, in general, **reason equationally** about effectful programs!

# Can we program with pure functions?

Yes! We've been doing it for the past 6 weeks.

Typically, a computation involving some state of type  $s$  and returning a result of type  $a$  can be expressed as a function:

$$s \rightarrow (s, a)$$

Rather than change the state, we return a new copy of the state.

## Efficiency?

All that copying might seem expensive, but by using tree data structures, we can usually reduce the cost to an  $\mathcal{O}(\log n)$  overhead.

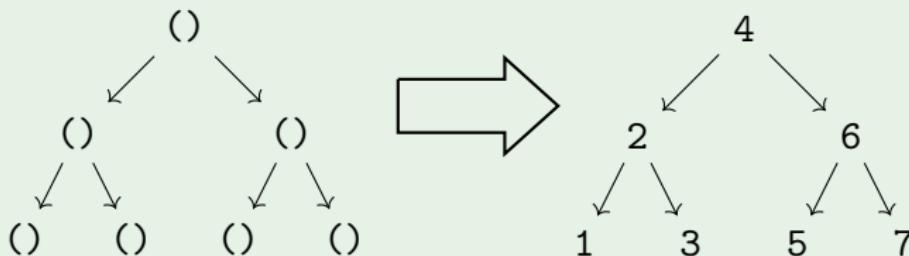
# State Passing

## Example (Labelling Nodes)

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

Given a tree, label each node with an ascending number in infix order:

```
label :: Tree () -> Tree Int
```



Let's use **monads** to simplify this!

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
○

# State Monads

```
newtype State s a = State (s -> (s, a))
```

## State Monad

```
get :: State s s
put :: s -> State s ()
modify :: (s -> s) -> State s ()
```

Here we use a **monadic** interface to simplify the passing of our state around, so that we don't need to manually plumb data around.

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
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# Effects

Sometimes we need side effects.

- We need to perform I/O, to communicate with the user or hardware.
- We might need effects for maximum efficiency.  
*(but usually internal effects are sufficient)*

## Haskell's approach

Pure by default. Effectful when necessary.

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
○

## The IO Type

A **procedure** that performs some side effects, returning a result of type a is written as `IO a`.

### World interpretation

`IO a` is an abstract type. But we can think of it as a function:

`RealWorld -> (RealWorld, a)`

(that's how it's implemented in GHC)

```
(>>=) :: IO a -> (a -> IO b) -> IO b
pure   :: a -> IO a
```

```
getChar :: IO Char
readLine :: IO String
putStrLn :: String -> IO ()
```

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
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## Infectious IO

We can convert pure values to impure procedures with `pure`:

```
pure :: a -> IO a
```

But we can't convert impure procedures to pure values:

```
???? :: IO a -> a
```

The only function that gets an `a` from an `IO a` is `>>=`:

```
(>>=) :: IO a -> (a -> IO b) -> IO b
```

But it returns an IO procedure as well.

### Conclusion

The moment you use an IO procedure in a function, IO shows up in the types, and you can't get rid of it!

If a function makes use of IO effects directly or indirectly, it will have IO in its type!

Effects  
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State Monad  
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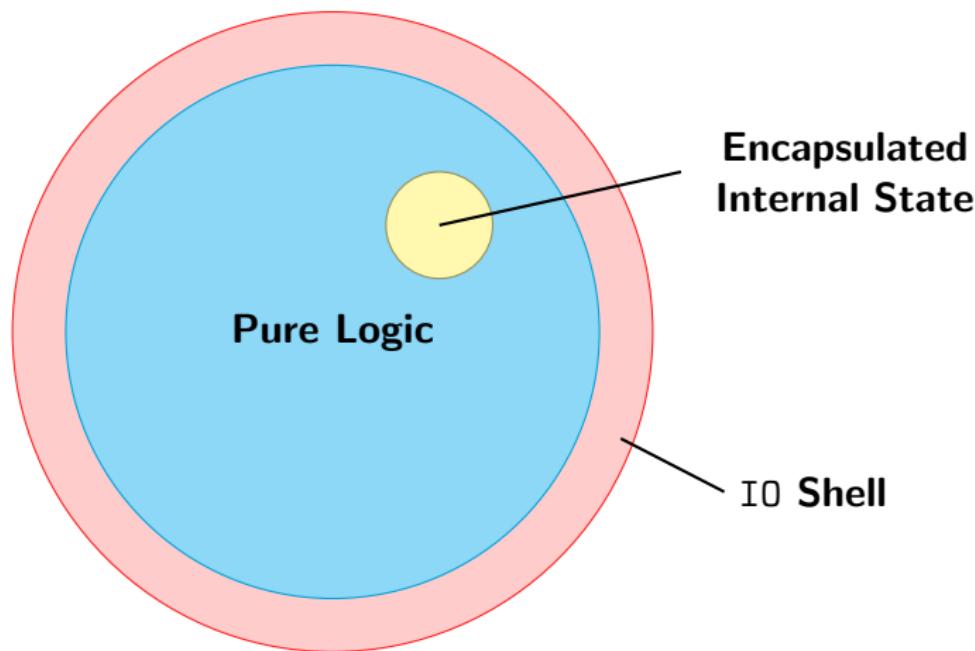
IO Monad  
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QuickChecking Monads  
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# Haskell Design Strategy

We ultimately “run” IO procedures by calling them from `main`:

```
main :: IO ()
```



Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
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# Examples

## Example (Triangles)

Given an input number  $n$ , print a triangle of  $*$  characters of base width  $n$ .

## Example (Maze Game)

Design a game that reads in a  $n \times n$  maze from a file. The player starts at position  $(0, 0)$  and must reach position  $(n - 1, n - 1)$  to win. The game accepts keyboard input to move the player around the maze.

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
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## Benefits of an IO Type

- Absence of effects makes type system more informative:
  - A type signatures captures **entire interface** of the function.
  - All **dependencies are explicit** in the form of data dependencies.
  - All **dependencies are typed**.
- It is easier to reason about pure code and it is easier to test:
  - Testing is local, doesn't require complex set-up and tear-down.
  - Reasoning is local, doesn't require state invariants.
  - Type checking leads to strong guarantees.

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
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## Mutable Variables

We can have honest-to-goodness mutability in Haskell, if we really need it, using `IORef`.

```
data IORef a
newIORef :: a -> IO (IORef a)
readIORef :: IORef a -> IO a
writeIORef :: IORef a -> a -> IO ()
```

### Example (Effectful Average)

Average a list of numbers using `IORefs`.

# Mutable Variables, Locally

Something like averaging a list of numbers doesn't require external effects, even if we use mutation internally.

```
data STRef s a
newSTRef :: a -> ST (STRef s a)
readSTRef :: STRef s a -> ST s a
writeSTRef :: STRef s a -> a -> ST s ()
runST :: (forall s. ST s a) -> a
```

The extra `s` parameter is called a **state thread**, that ensures that mutable variables don't leak outside of the ST computation.

## Note

The ST Monad is not assessable in this course, but it is useful sometimes in Haskell programming.

Effects  
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State Monad  
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IO Monad  
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QuickChecking Monads  
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## QuickChecking Monads

QuickCheck lets us test IO (and ST) using this special **property monad** interface:

```
monadicIO  :: PropertyM IO () -> Property
pre        :: Bool -> PropertyM IO ()
assert     :: Bool -> PropertyM IO ()
run        :: IO a -> PropertyM IO a
```

### Example (Testing average)

Let's test that our IO average function works like the non-effectful one.

### Example (Testing gfactor)

Let's test that the GNU factor program works correctly!

# COMP3141

Software System Design and Implementation

## Effects and IO Monad Practice

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Term 2 2019

## Recall: The IO Type

A **procedure** that performs some side effects, returning a result of type a is written as `IO a`.

### World interpretation

`IO a` is an abstract type. But we can think of it as a function:

`RealWorld -> (RealWorld, a)`

(that's how it's implemented in GHC)

```
(>>=) :: IO a -> (a -> IO b) -> IO b  
pure   :: a -> IO a
```

```
getChar :: IO Char  
readLine :: IO String  
putStrLn :: String -> IO ()
```

## QuickChecking Monads

QuickCheck lets us test IO (and ST) using this special **property monad** interface:

```
monadicIO  :: PropertyM IO () -> Property
pre        :: Bool -> PropertyM IO ()
assert     :: Bool -> PropertyM IO ()
run        :: IO a -> PropertyM IO a
```

### Example (Testing hash)

Let's test that our IO password hash function works like GHC's non-effectful one.

- This implementation is functionally correct but is it secure?
- Does functional correctness imply security?
- Could hash still be identity? Is it a good hash function?

## Recall: State Monads

```
newtype State s a = State (s -> (s, a))
```

### State Monad

```
get :: State s s
put :: s -> State s ()
modify :: (s -> s) -> State s ()
```

Here we use a **monadic** interface to simplify the passing of our state around, so that we don't need to manually plumb data around.

# Fibonacci Example

## Example (Testing Fibonacci)

Let's test that our stateful Fibonacci function works like the pure one.

- Does the performance of the abstract model matter when testing?
- But shouldn't our abstract model be as abstract as possible?
- This is a cost that testing incurs as opposed to formal verification

Static Assurance  
oooo

Phantom Types  
oooooo

GADTs  
oooooooooooo

Type Families  
oooo

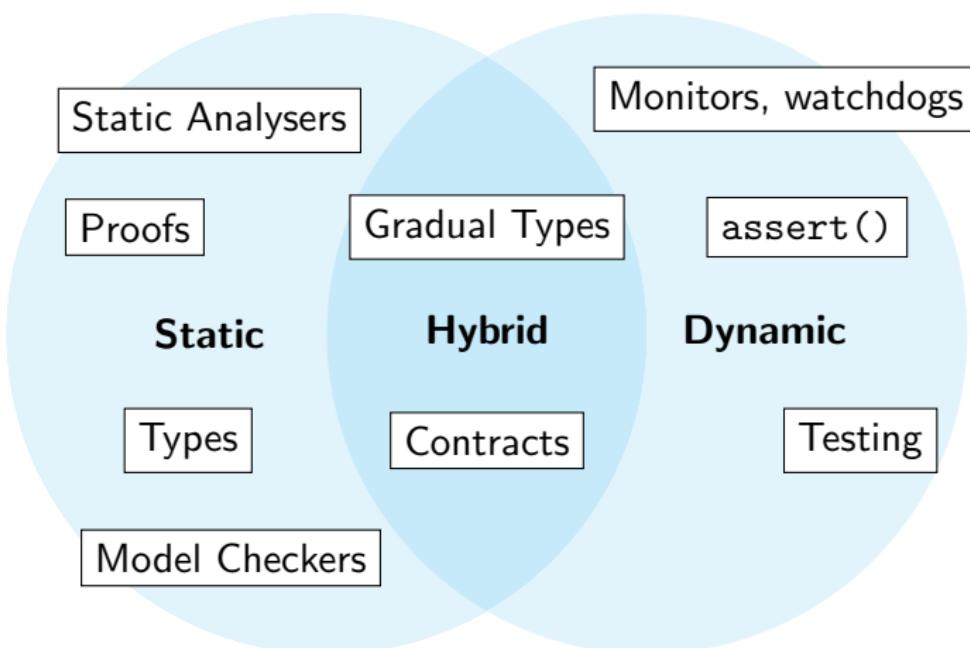
# COMP3141

## Software System Design and Implementation

### Static Assurance with Types

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Term 2 2019

# Methods of Assurance



Static means of assurance analyse a program **without running it**.

# Static vs. Dynamic

- Static checks can be **exhaustive**.

## Exhaustivity

An exhaustive check is a check that is able to analyse all possible executions of a program.

- **However**, some properties cannot be checked statically in general (**halting problem**), or are intractable to feasibly check statically (**state space explosion**).
- Dynamic checks cannot be exhaustive, but can be used to check some properties where static methods are unsuitable.

# Compiler Integration

Most static and all dynamic methods of assurance are **not** integrated into the compilation process.

- You can compile and run your program even if it fails tests.
- You can change your program to diverge from your model checker model.
- Your proofs can diverge from your implementation.

## Types

Because types **are** integrated into the compiler, they cannot diverge from the source code. This means that type signatures are a kind of **machine-checked documentation** for your code.

# Types

Types are the **most widely used** kind of formal verification in programming today.

- They are checked automatically by the compiler.
- They can be extended to encompass properties and proof systems with very high expressivity (covered next week).
- They are an **exhaustive** analysis.

This week, we'll look at techniques to encode various correctness conditions **inside Haskell's type system**.



# Phantom Types

## Definition

A type parameter is *phantom* if it does not appear in the right hand side of the type definition.

```
newtype Size x = S Int
```

Lets examine each one of the following use cases:

- We can use this parameter to track what *data invariants* have been established about a value.
- We can use this parameter to track information about the representation (e.g. units of measure).
- We can use this parameter to enforce an *ordering* of operations performed on these values (*type state*).

# Validation

```
data UG -- empty type
data PG
data StudentID x = SID Int
```

We can define a **smart constructor** that specialises the type parameter:

```
sid :: Int -> Either (StudentID UG)
                           (StudentID PG)
```

(Recalling the following definition of Either)

```
data Either a b = Left a | Right b
```

And then define functions:

```
enrolInCOMP3141 :: StudentID UG -> IO ()
lookupTranscript :: StudentID x -> IO String
```

# Units of Measure

In 1999, software confusing units of measure (pounds and newtons) caused a mars orbiter to burn up on atmospheric entry.

```
data Kilometres
data Miles
data Value x = U Int
sydneyToMelbourne = (U 877 :: Value Kilometres)
losAngelesToSanFran = (U 383 :: Value Miles)
```

In addition to tagging values, we can also enforce constraints on units:

```
data Square a
area :: Value m -> Value m -> Value (Square m)
area (U x) (U y) = U (x * y)
```

Note the arguments to area must have the same units.

# Type State

## Example

A Socket can either be ready to receive data, or busy. If the socket is busy, the user must first use the `wait` operation, which blocks until the socket is ready. If the socket is ready, the user can use the `send` operation to send string data, which will make the socket busy again.

```
data Busy
data Ready
newtype Socket s = Socket ...

wait :: Socket Busy -> IO (Socket Ready)

send :: Socket Ready -> String -> IO (Socket Busy)
```

What assumptions are we making here?

# Linearity and Type State

The previous code assumed that we didn't re-use old Sockets:

```
send2 :: Socket Ready -> String -> String
      -> IO (Socket Busy)
send2 s x y = do s' <- send s x
                  s'' <- wait s'
                  s''' <- send s'' y
                  pure s'''
```

But we can just re-use old values to send without waiting:

```
send2' s x y = do _ <- send s x
                     s' <- send s y
                     pure s'
```

*Linear type* systems  
can solve this, but  
not in Haskell (yet).

# Datatype Promotion

```
data UG
data PG
data StudentID x = SID Int
```

Defining empty data types for our tags is **untyped**. We can have `StudentID UG`, but also `StudentID String`.

## Recall

Haskell types themselves have types, called **kinds**. Can we make the kind of our tag types more precise than `*?`

The DataKinds language extension lets us use data types as kinds:

```
{-# LANGUAGE DataKinds, KindSignatures #-}
data Stream = UG | PG
data StudentID (x :: Stream) = SID Int
-- rest as before
```

# Motivation: Evaluation

```
data Expr = BConst Bool
          | IConst Int
          | Times Expr Expr
          | Less Expr Expr
          | And Expr Expr
          | If Expr Expr Expr
data Value = BVal Bool | IVal Int
```

## Example

Define an expression evaluator:

```
eval :: Expr -> Value
```

## Motivation: Partiality

Unfortunately the eval function is **partial**, undefined for input expressions that are not well-typed, like:

`And (ICons 3) (BConst True)`

### Recall

With any partial function, we can make it total by either **expanding** the co-domain (e.g. with a Maybe type), or **constraining** the domain.

Can we use phantom types to constrain the domain of eval to only accept well-typed expressions?

## Attempt: Phantom Types

Let's try adding a phantom parameter to Expr, and defining typed constructors with precise types:

```
data Expr t = ...
bConst :: Bool -> Expr Bool
bConst = BConst
iConst :: Int -> Expr Int
iConst = IConst
times :: Expr Int -> Expr Int -> Expr Int
times = Times
less :: Expr Int -> Expr Int -> Expr Bool
less = Less
and :: Expr Bool -> Expr Bool -> Expr Bool
and = And
if' :: Expr Bool -> Expr a -> Expr a -> Expr a
if' = If
```

# Attempt: Phantom Types

This makes invalid expressions into type errors (yay!):

```
-- Couldn't match Int and Bool  
and (iCons 3) (bConst True)
```

How about our eval function? What should its type be now?

```
eval :: Expr t -> t
```

## Bad News

Inside eval, the Haskell type checker cannot be sure that we used our typed constructors, so in e.g. the IConst case:

```
eval :: Expr t -> t  
eval (IConst i) = i -- type error
```

We are unable to tell that the type t is definitely Int.

Phantom types aren't strong enough!

# GADTs

Generalised Algebraic Datatypes (*GADTs*) is an extension to Haskell that, among other things, allows data types to be specified by writing the types of their constructors:

```
{-# LANGUAGE GADTs, KindSignatures #-}
-- Unary natural numbers, e.g. 3 is S (S (S Z))
data Nat = Z | S Nat
-- is the same as
data Nat :: * where
  Z :: Nat
  S :: Nat -> Nat
```

When combined with the *type indexing* trick of phantom types, this becomes very powerful!

# Expressions as a GADT

```
data Expr :: * -> * where
  BConst :: Bool -> Expr Bool
  IConst :: Int -> Expr Int
  Times :: Expr Int -> Expr Int -> Expr Int
  Less :: Expr Int -> Expr Int -> Expr Bool
  And :: Expr Bool -> Expr Bool -> Expr Bool
  If   :: Expr Bool -> Expr a -> Expr a -> Expr a
```

## Observation

There is now only *one* set of precisely-typed constructors.

Inside eval now, the Haskell type checker accepts our previously problematic case:

```
eval :: Expr t -> t
eval (IConst i) = i -- OK now
```

GHC now knows that if we have IConst, the type t must be Int.

# Lists

We could define our own list type using GADT syntax as follows:

```
data List (a :: *) :: * where
  Nil  :: List a
  Cons :: a -> List a -> List a
```

But, if we define head (hd) and tail (tl) functions, they're **partial** (boo!):

```
hd (Cons x xs) = x
tl (Cons x xs) = xs
```

We will constrain the domain of these functions by tracking the **length** of the list **on the type level**.

# Vectors

As before, define a natural number kind to use on the type level:

```
data Nat = Z | S Nat
```

Now our length-indexed list can be defined, called a Vec:

```
data Vec (a :: *) :: Nat -> * where
  Nil :: Vec a Z
  Cons :: a -> Vec a n -> Vec a (S n)
```

Now hd and tl can be total:

```
hd :: Vec a (S n) -> a
hd (Cons x xs) = x
tl :: Vec a (S n) -> Vec a n
tl (Cons x xs) = xs
```

# Vectors, continued

Our `map` for vectors is as follows:

```
mapVec :: (a -> b) -> Vec a n -> Vec b n
mapVec f Nil = Nil
mapVec f (Cons x xs) = Cons (f x) (mapVec f xs)
```

## Properties

Using this type, it's impossible to write a `mapVec` function that changes the length of the vector.

**Properties are verified by the compiler!**

## Tradeoffs

The benefits of this extra static checking are obvious, however:

- It can be difficult to convince the Haskell type checker that your code is correct, even when it is.
- Type-level encodings can make types more verbose and programs harder to understand.
- Sometimes excessively detailed types can make type-checking very slow, hindering productivity.

## Pragmatism

We should use type-based encodings only when the assurance advantages outweigh the clarity disadvantages.

The typical use case for these richly-typed structures is to eliminate **partial functions** from our code base.

If we never use partial list functions, length-indexed vectors are not particularly useful.

# Appending Vectors

## Example (Problem)

```
appendV :: Vec a m -> Vec a n -> Vec a ???
```

We want to write  $m + n$  in the  $???$  above, but we do not have addition defined for kind Nat.

We can define a normal Haskell function easily enough:

```
plus :: Nat -> Nat -> Nat
plus Z y = y
plus (S x) y = S (plus x y)
```

This function is not applicable to **type-level** Nats, though.  
⇒ we need a **type level function**.

# Type Families

Type level functions, also called *type families*, are defined in Haskell like so:

```
{-# LANGUAGE TypeFamilies #-}  
type family Plus (x :: Nat) (y :: Nat) :: Nat where  
    Plus Z      y = y  
    Plus (S x) y = S (Plus x y)
```

We can use our type family to define appendV:

```
appendV :: Vec a m -> Vec a n -> Vec a (Plus m n)  
appendV Nil          ys = ys  
appendV (Cons x xs) ys = Cons x (appendV xs ys)
```

# Recursion

If we had implemented Plus by recursing on the second argument instead of the first:

```
{-# LANGUAGE TypeFamilies #-}  
type family Plus' (x :: Nat) (y :: Nat) :: Nat where  
  Plus' x Z      = x  
  Plus' x (S y) = S (Plus' x y)
```

Then our appendV code would not type check.

```
appendV :: Vec a m -> Vec a n -> Vec a (Plus' m n)  
appendV Nil          ys = ys  
appendV (Cons x xs) ys = Cons x (appendV xs ys)
```

Why?

## Answer

Consider the Nil case. We know  $m = Z$ , and must show that our desired return type  $\text{Plus}' Z n$  equals our given return type  $n$ , but that fact is not immediately apparent from the equations.

# Type-driven development

- This lecture is only a taste of the full power of type-based specifications.
- Languages supporting **dependent types** (Idris, Agda) completely merge the type and value level languages, and support machine-checked proofs about programs.
- Haskell is also gaining more of these typing features all the time.

**Next week:** Fancy theory about types!

- Deep connections between types, logic and proof.
- Algebraic type structure for generic algorithms and refactoring.
- Using polymorphic types to infer properties for free.

# COMP3141

## Software System Design and Implementation

### GADTs Practice

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## Recall: GADTs

Generalised Algebraic Datatypes (*GADTs*) is an extension to Haskell that, among other things, allows data types to be specified by writing the types of their constructors:

```
{-# LANGUAGE GADTs, KindSignatures #-}
-- Unary natural numbers, e.g. 3 is S (S (S Z))
data Nat = Z | S Nat
-- is the same as
data Nat :: * where
  Z :: Nat
  S :: Nat -> Nat
```

## The printf function

Consider the well known C function printf:

```
printf("Hello %s You are %d years old!", "Nina", 22)
```

In C, the type (and number) of parameters passed to this function are dependent on the first parameter (the format string).

## The printf function

To do a similar thing in Haskell, we would like to use a richer type that allows us to define a function whose subsequent parameter is determined by the first.

```
data Format :: * -> * where
    End :: Format ()                                -- Empty format
    Str :: Format t -> Format (String, t)          -- %s
    Dec :: Format t -> Format (Int, t)              -- %d
    L :: String -> Format t -> Format t           -- literal strings
deriving instance Show (Format ts)
-- just like deriving (Show) for normal data types.
```

Our format strings is indexed by a tuple type containing all of the types of the %directives used.

# The printf function

"Hello %s You are %d years old!"

is written:

```
L "Hello" $ Str $ L " You are "
$ Dec $ L " years old!" End
```

# The printf function

```
printf :: Format ts -> ts -> IO ()  
printf End () =  
    pure () -- type is ()  
printf (Str fmt) (s,ts) =  
    do putStrLn s; printf fmt ts -- type is (String, ...)  
printf (Dec fmt) (i,ts) =  
    do putStrLn (show i); printf fmt ts -- type is (Int, ...)  
printf (L s fmt) ts      =  
    do putStrLn s; printf fmt ts
```

# Vectors

Define a natural number kind to use on the type level:

```
data Nat = Z | S Nat
```

Our length-indexed list can be defined, called a Vec:

```
data Vec (a :: *) :: Nat -> * where
  Nil :: Vec a Z
  Cons :: a -> Vec a n -> Vec a (S n)
```

The functions hd and tl can be total:

```
hd :: Vec a (S n) -> a
hd (Cons x xs) = x
tl :: Vec a (S n) -> Vec a n
tl (Cons x xs) = xs
```

# Vectors, continued

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We can define a normal Haskell function easily enough:

```
plus :: Nat -> Nat -> Nat
plus Z y = y
plus (S x) y = S (plus x y)
```

This function is not applicable to **type-level** Nats, though.  
⇒ we need a **type level function**.

# Type Families

Type level functions, also called *type families*, are defined in Haskell like so:

```
{-# LANGUAGE TypeFamilies #-}  
type family Plus (x :: Nat) (y :: Nat) :: Nat where  
    Plus Z      y = y  
    Plus (S x) y = S (Plus x y)
```

We can use our type family to define appendV:

```
appendV :: Vec a m -> Vec a n -> Vec a (Plus m n)  
appendV Nil          ys = ys  
appendV (Cons x xs) ys = Cons x (appendV xs ys)
```

# Concatenating Vectors

## Example (Problem)

```
concatV :: Vec (Vec a m) n -> Vec a ???
```

We want to write  $m * n$  in the  $???$  above, but we do not have times defined for kind Nat.

```
{-# LANGUAGE TypeFamilies #-}  
type family Times (a :: Nat) (b :: Nat) :: Nat where  
  Times Z n = Z  
  Times (S m) n = Plus n (Times m n)
```

We can use our type family to define concatV:

```
concatV :: Vec (Vec a m) n -> Vec a (Times n m)  
concatV Nil = Nil  
concatV (Cons v vs) = v `appendV` concatV vs
```

# Filtering Vectors

## Example (Problem)

```
filterV :: (a -> Bool) -> Vec a n -> Vec a ???
```

What is the size of the result of filter?

# Filtering Vectors

## Example (Problem)

```
filterV :: (a -> Bool) -> Vec a n -> [a]
```

We do not know the size of the result.

We can use our type family to define concatV:

```
filterV :: (a -> Bool) -> Vec a n -> [a]
filterV p Nil = []
filterV p (Cons x xs)
  | p x  = x : filterV p xs
  | otherwise = filterV p xs
```

## Recap: Logic

•

## Typed Lambda Calculus

A horizontal row of 20 small circles, each containing a dot.

## Algebraic Type Isomorphism

oo

## Polymorphism and Parametricity

○○○○○○○○○○

# **COMP3141**

## Software System Design and Implementation

## Theory of Types

Liam O'Connor  
CSE, UNSW (and data61)  
Term 2 2019

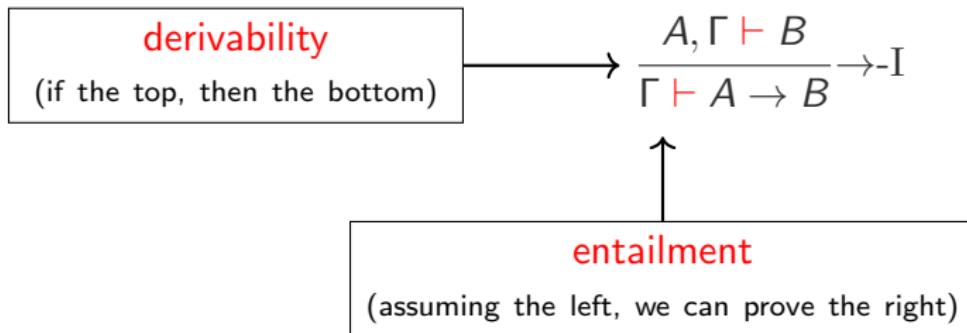
# Natural Deduction

# Logic

We can specify a logical system as a *deductive system* by providing a set of **rules** and **axioms** that describe how to prove various connectives.

Each connective typically has *introduction* and *elimination* rules.

For example, to prove an implication  $A \rightarrow B$  holds, we must show that  $B$  holds assuming  $A$ . This introduction rule is written as:



## More rules

Implication also has an elimination rule, that is also called *modus ponens*:

$$\frac{\Gamma \vdash A \rightarrow B \quad \Gamma \vdash A}{\Gamma \vdash B} \rightarrow\text{-E}$$

Conjunction (*and*) has an introduction rule that follows our intuition:

$$\frac{\Gamma \vdash A \quad \Gamma \vdash B}{\Gamma \vdash A \wedge B} \wedge\text{-I}$$

It has **two** elimination rules:

$$\frac{\Gamma \vdash A \wedge B}{\Gamma \vdash A} \wedge\text{-E}_1 \quad \frac{\Gamma \vdash A \wedge B}{\Gamma \vdash B} \wedge\text{-E}_2$$

## More rules

Disjunction (or) has two introduction rules:

$$\frac{\Gamma \vdash A}{\Gamma \vdash A \vee B} \vee\text{-I}_1 \quad \frac{\Gamma \vdash B}{\Gamma \vdash A \vee B} \vee\text{-I}_2$$

Disjunction elimination is a little unusual:

$$\frac{\Gamma \vdash A \vee B \quad A, \Gamma \vdash P \quad B, \Gamma \vdash P}{\Gamma \vdash P} \vee\text{-E}$$

The true literal, written  $\top$ , has only an introduction:

上 T

And false, written  $\perp$ , has just elimination (*ex falso quodlibet*):

$$\frac{\Gamma \vdash \perp}{\Gamma \vdash P}$$

## Example Proofs

## Example

Prove:

- $A \wedge B \rightarrow B \wedge A$
  - $A \vee \perp \rightarrow A$

What would **negation** be equivalent to?

Typically we just define

$$\neg A \equiv (A \rightarrow \perp)$$

## Example

Prove:

- $A \rightarrow (\neg\neg A)$
  - $(\neg\neg A) \rightarrow A$  We get stuck here!

# Constructive Logic

The logic we have expressed so far does **not** admit the law of the excluded middle:

$$P \vee \neg P$$

Or the equivalent double negation elimination:

$$(\neg \neg P) \rightarrow P$$

This is because it is a *constructive* logic that does not allow us to do proof by contradiction.

# Boiling Haskell Down

The theoretical properties we will describe also apply to Haskell, but we need a smaller language for demonstration purposes.

- No user-defined types, just a small set of built-in types.
  - No polymorphism (type variables)
  - Just lambdas ( $\lambda x.e$ ) to define functions or bind variables.

This language is a very minimal functional language, called the **simply typed lambda calculus**, originally due to Alonzo Church.

Our small set of built-in types are intended to be enough to express most of the data types we would otherwise define. We are going to use logical inference rules to specify how expressions are given types (*typing rules*).

## Function Types

We create values of a function type  $A \rightarrow B$  using lambda expressions:

$$\frac{x :: A, \Gamma \vdash e :: B}{\Gamma \vdash \lambda x. e :: A \rightarrow B}$$

The typing rule for function application is as follows:

$$\frac{\Gamma \vdash e_1 :: A \rightarrow B \quad \Gamma \vdash e_2 :: A}{\Gamma \vdash e_1 \ e_2 :: B}$$

What other types would be needed?

## Composite Data Types

In addition to functions, most programming languages feature ways to *compose* types together to produce new types, such as:

# Classes

# Tuples

# Unions

## Records

## Combining values conjunctively

We want to store two things in one value.

(might want to use non-compact slides for this one)

```
Has  
type Point  
midpoint ()
```

## Product Types

For simply typed lambda calculus, we will accomplish this with tuples, also called *product types*.

$$(A, B)$$

We won't have type declarations, named fields or anything like that. More than two values can be combined by nesting products, for example a three dimensional vector:

$(\text{Int}, (\text{Int}, \text{Int}))$

# Constructors and Eliminators

We can **construct** a product type the same as Haskell tuples:

$$\frac{\Gamma \vdash e_1 :: A \quad \Gamma \vdash e_2 :: B}{\Gamma \vdash (e_1, e_2) :: (A, B)}$$

The only way to extract each component of the product is to use the `fst` and `snd` eliminators:

$$\frac{\Gamma \vdash e :: (A, B)}{\Gamma \vdash \text{fst } e :: A} \quad \frac{\Gamma \vdash e :: (A, B)}{\Gamma \vdash \text{snd } e :: B}$$

## Unit Types

Currently, we have no way to express a type with just **one** value. This may seem useless at first, but it becomes useful in combination with other types.

We'll introduce the **unit** type from Haskell, written `()`, which has exactly one inhabitant, also written `()`:

Γ ⊢ () :: ()

## Disjunctive Composition

We can't, with the types we have, express a type with exactly **three** values.

### Example (Trivalued type)

```
data TrafficLight = Red | Amber | Green
```

In general we want to express data that can be **one** of multiple **alternatives**, that contain different bits of data.

## Example (More elaborate alternatives)

```
type Length = Int
type Angle = Int
data Shape = Rect Length Length
            | Circle Length | Point
            | Triangle Angle Length Length
```

This is awkward in many languages. In Java we'd have to use inheritance. In C we'd have to use unions.

# Sum Types

We'll build in the Haskell Either type to express the possibility that data may be one of two forms.

Either A B

These types are also called *sum types*.

Our TrafficLight type can be expressed (grotesquely) as a sum of units:

`TrafficLight`  $\simeq$  `Either () (Either () ())`

# Constructors and Eliminators for Sums

To make a value of type Either  $A$   $B$ , we invoke one of the two constructors:

$$\frac{\Gamma \vdash e :: A}{\Gamma \vdash \text{Left } e :: \text{Either } A B} \quad \frac{\Gamma \vdash e :: B}{\Gamma \vdash \text{Right } e :: \text{Either } A B}$$

We can branch based on which alternative is used using pattern matching:

$$\frac{\Gamma \vdash e :: \text{Either } A B \quad x :: A, \Gamma \vdash e_1 :: P \quad y :: B, \Gamma \vdash e_2 :: P}{\Gamma \vdash (\text{case } e \text{ of Left } x \rightarrow e_1; \text{Right } y \rightarrow e_2) :: P}$$

# Examples

## Example (Traffic Lights)

Our traffic light type has three values as required:

$$\text{TrafficLight} \simeq \text{Either} () (\text{Either} () ())$$

$$\text{Red} \simeq \text{Left} ()$$

$$\text{Amber} \simeq \text{Right} (\text{Left} ())$$

$$\text{Green} \simeq \text{Right} (\text{Right} (\text{Left} ()))$$

## The Empty Type

We add another type, called Void, that has no inhabitants.  
Because it is empty, there is no way to construct it.  
We do have a way to eliminate it, however:

$$\frac{\Gamma \vdash e :: \text{Void}}{\Gamma \vdash \text{absurd } e :: P}$$

If I have a variable of the **empty** type in scope, we must be looking at an expression that will **never** be evaluated. Therefore, we can assign any type we like to this expression, because it will never be executed.

# Gathering Rules

$$\frac{\Gamma \vdash e :: \text{Void}}{\Gamma \vdash \text{absurd } e :: P} \quad \frac{}{\Gamma \vdash () :: ()}$$
$$\frac{\Gamma \vdash e :: A}{\Gamma \vdash \text{Left } e :: \text{Either } A B} \quad \frac{\Gamma \vdash e :: B}{\Gamma \vdash \text{Right } e :: \text{Either } A B}$$
$$\frac{\Gamma \vdash e :: \text{Either } A B \quad x :: A, \Gamma \vdash e_1 :: P \quad y :: B, \Gamma \vdash e_2 :: P}{\Gamma \vdash (\text{case } e \text{ of Left } x \rightarrow e_1; \text{Right } y \rightarrow e_2) :: P}$$
$$\frac{\Gamma \vdash e_1 :: A \quad \Gamma \vdash e_2 :: B}{\Gamma \vdash (e_1, e_2) :: (A, B)} \quad \frac{\Gamma \vdash e :: (A, B)}{\Gamma \vdash \text{fst } e :: A} \quad \frac{\Gamma \vdash e :: (A, B)}{\Gamma \vdash \text{snd } e :: B}$$
$$\frac{\Gamma \vdash e_1 :: A \rightarrow B \quad \Gamma \vdash e_2 :: A}{\Gamma \vdash e_1 \ e_2 :: B} \quad \frac{x :: A, \Gamma \vdash e :: B}{\Gamma \vdash \lambda x. \ e :: A \rightarrow B}$$

# Removing Terms. . .

$$\frac{\Gamma \vdash \text{Void}}{\Gamma \vdash P} \quad \frac{}{\Gamma \vdash ()}$$
$$\frac{\Gamma \vdash A}{\Gamma \vdash \text{Either } A \ B} \quad \frac{\Gamma \vdash B}{\Gamma \vdash \text{Either } A \ B}$$
$$\frac{\Gamma \vdash \text{Either } A \ B \quad A, \Gamma \vdash P \quad B, \Gamma \vdash P}{\Gamma \vdash P}$$
$$\frac{\Gamma \vdash A \quad \Gamma \vdash B}{\Gamma \vdash (A, B)} \quad \frac{\Gamma \vdash (A, B)}{\Gamma \vdash A} \quad \frac{\Gamma \vdash (A, B)}{\Gamma \vdash B}$$
$$\frac{\Gamma \vdash A \rightarrow B \quad \Gamma \vdash A}{\Gamma \vdash B} \quad \frac{A, \Gamma \vdash B}{\Gamma \vdash A \rightarrow B}$$

This looks exactly like **constructive logic!**

If we can construct a **program** of a certain **type**, we have also created a **proof** of a certain **proposition**.

# The Curry-Howard Correspondence

This correspondence goes by many names, but is usually attributed to **Haskell Curry** and **William Howard**.

It is a *very deep* result:

Programming	Logic
Types	Propositions
Programs	Proofs
Evaluation	Proof Simplification

It turns out, no matter what logic you want to define, there is always a corresponding  $\lambda$ -calculus, and vice versa.

Typed $\lambda$ -Calculus Continuations Monads Linear Types, Session Types Region Types	Constructive Logic Classical Logic Modal Logic Linear Logic Separation Logic
---	--

## Examples

### Example (Commutativity of Conjunction)

*andComm :: (A, B) → (B, A)*

*andComm p = (snd p, fst p)*

This proves  $A \wedge B \rightarrow B \wedge A$ .

### Example (Transitivity of Implication)

*transitive :: (A → B) → (B → C) → (A → C)*

*transitive f g x = g (f x)*

Transitivity of implication is just **function composition**.

# Translating

We can translate logical connectives to types and back:

Tuples	Conjunction ( $\wedge$ )
Either	Disjunction ( $\vee$ )
Functions	Implication
()	True
Void	False

We can also translate our *equational reasoning* on programs into *proof simplification* on proofs!

# Proof Simplification

Assuming  $A \wedge B$ , we want to prove  $B \wedge A$ .

We have this unpleasant proof:

$$\frac{\begin{array}{c} A \wedge B \\ \hline A \end{array} \quad \begin{array}{c} A \wedge B \\ \hline A \end{array}}{\begin{array}{c} A \wedge B \\ \hline A \wedge A \\ \hline A \end{array}}$$
$$\frac{\begin{array}{c} A \wedge B \\ \hline B \end{array}}{B \wedge A}$$

# Proof Simplification

Translating to types, we get:

Assuming  $x :: (A, B)$ , we want to construct  $(B, A)$ .

$$\begin{array}{c}
 \dfrac{x :: (A, B)}{\text{fst } x :: A} \qquad \dfrac{x :: (A, B)}{\text{fst } x :: A} \\
 \hline
 \dfrac{x :: (A, B)}{\text{snd } x :: B} \qquad \dfrac{\text{fst } x, \text{fst } x :: (A, A)}{\text{snd } (\text{fst } x, \text{fst } x) :: A} \\
 \hline
 \dfrac{}{(\text{snd } x, \text{snd } (\text{fst } x, \text{fst } x)) :: (B, A)}
 \end{array}$$

We know that

$$(\text{snd } x, \text{snd } (\text{fst } x, \text{fst } x)) = (\text{snd } x, \text{fst } x)$$

Lets apply this simplification to our proof!

# Proof Simplification

Assuming  $x :: (A, B)$ , we want to construct  $(B, A)$ .

$$\frac{\begin{array}{c} x :: (A, B) \\ \hline \text{snd } x :: B \end{array}}{\frac{x :: (A, B)}{\text{fst } x :: A}} \quad \frac{\text{snd } x :: B \quad \text{fst } x :: A}{( \text{snd } x, \text{fst } x ) :: (B, A)}$$

Back to logic:

$$\frac{\begin{array}{c} A \wedge B \\ \hline B \end{array} \quad \frac{A \wedge B}{A}}{B \wedge A}$$

# Applications

As mentioned before, in **dependently typed languages** such as Agda and Idris, the distinction between value-level and type-level languages is removed, allowing us to refer to our program in types (i.e. propositions) and then construct programs of those types (i.e. proofs).

## Peano Arithmetic

If there's time, Liam will demo how to prove some basic facts of natural numbers in Agda, a dependently typed language.

Generally, dependent types allow us to use rich types not just for programming, but also for verification via the Curry-Howard correspondence.

## Caveats

All functions we define have to be **total and terminating**.

Otherwise we get an ***inconsistent*** logic that lets us prove false things:

$$\text{proof}_1 :: P = NP$$

$$\text{proof}_1 = \text{proof}_1$$

$$\text{proof}_2 :: P \neq NP$$

$$\text{proof}_2 = \text{proof}_2$$

Most common calculi correspond to **constructive** logic, not **classical** ones, so principles like the **law of excluded middle** or **double negation elimination** do **not** hold:

$$\neg\neg P \rightarrow P$$

## Semiring Structure

These types we have defined form an algebraic structure called a *commutative semiring*.

Laws for Either and Void:

- Associativity:

$$\text{Either}(\text{Either } A \ B) \ C \simeq \text{Either } A (\text{Either } B \ C)$$

- Identity:  $\text{Either Void } A \simeq A$

- Commutativity:  $\text{Either } A \ B \simeq \text{Either } B \ A$

Laws for tuples and 1

- Associativity:  $((A, B), C) \simeq (A, (B, C))$

- Identity:  $((), A) \simeq A$

- Commutativity:  $(A, B) \simeq (B, A)$

Combining the two:

- Distributivity:  $(A, \text{Either } B \ C) \simeq \text{Either } (A, B) (A, C)$

- Absorption:  $(\text{Void}, A) \simeq \text{Void}$

What does  $\simeq$  mean here? It's more than logical equivalence.

## Isomorphism

Two types  $A$  and  $B$  are *isomorphic*, written  $A \simeq B$ , if there exists a *bijection* between them. This means that for each value in  $A$  we can find a unique value in  $B$  and vice versa.

### Example (Refactoring)

We can use this reasoning to simplify type definitions. For example:

```
data Switch = On Name Int  
             | Off Name
```

Can be simplified to the isomorphic (Name, Maybe Int).

### Generic Programming

Representing data types generically as sums and products is the foundation for *generic programming* libraries such as GHC generics. This allows us to define algorithms that work on arbitrary data structures.

# Type Quantifiers

Consider the type of `fst`:

`fst :: (a,b) -> a`

This can be written more verbosely as:

`fst :: forall a b. (a,b) -> a`

Or, in a more mathematical notation:

`fst :: ∀a b. (a, b) → a`

This kind of quantification over type variables is called **parametric polymorphism** or just **polymorphism** for short.

(It's also called **generics** in some languages, but this terminology is bad)

What is the analogue of  $\forall$  in logic? (via Curry-Howard)?

# Curry-Howard

The type quantifier  $\forall$  corresponds to a universal quantifier  $\forall$ , but it is **not** the same as the  $\forall$  from first-order logic. What's the difference?

First-order logic quantifiers range over a set of *individuals* or values, for example the natural numbers:

$$\forall x, x + 1 > x$$

These quantifiers range over **propositions** (types) themselves. It is analogous to ***second-order logic***, not first-order:

$$\forall A. \forall B. A \wedge B \rightarrow B \wedge A$$

$$\forall A. \forall B. (A, B) \rightarrow (B, A)$$

The first-order quantifier has a type-theoretic analogue too (type indices), but this is not nearly as common as polymorphism.

# Generality

If we need a function of type  $\text{Int} \rightarrow \text{Int}$ , a polymorphic function of type  $\forall a. a \rightarrow a$  will do just fine, we can just instantiate the type variable to  $\text{Int}$ . But the reverse is not true. This gives rise to an ordering.

## Generality

A type  $A$  is *more general* than a type  $B$ , often written  $A \sqsubseteq B$ , if type variables in  $A$  can be instantiated to give the type  $B$ .

## Example (Functions)

$$\text{Int} \rightarrow \text{Int} \quad \equiv \quad \forall z. z \rightarrow z \quad \equiv \quad \forall x y. x \rightarrow y \quad \equiv \quad \forall a. a$$

## Constraining Implementations

How many possible total, terminating implementations are there of a function of the following type?

$\text{Int} \rightarrow \text{Int}$

How about this type?

$$\forall a. \ a \rightarrow a$$

Polymorphic type signatures constrain implementations.

# Parametricity

## Definition

The principle of **parametricity** states that the result of polymorphic functions cannot depend on **values** of an abstracted type.

More formally, suppose I have a polymorphic function  $g$  that is polymorphic on type  $a$ . If run any arbitrary function  $f :: a \rightarrow a$  on all the  $a$  values in the input of  $g$ , that will give the same results as running  $g$  first, then  $f$  on all the  $a$  values of the output.

## Example

*foo* ::  $\forall a. [a] \rightarrow [a]$

We know that **every** element of the output occurs in the input.  
 The parametricity theorem we get is, for all  $f$ :

$$foo \circ (map\ f) = (map\ f) \circ foo$$

## More Examples

*head* ::  $\forall a. [a] \rightarrow a$

## What's the parametricity theorems?

## Example (Answer)

For any  $f$ :

$$f \ (\text{head } \ell) = \text{head } (\text{map } f \ \ell)$$

## More Examples

(++) ::  $\forall a. [a] \rightarrow [a] \rightarrow [a]$

## What's the parametricity theorem?

## Example (Answer)

$$\text{map } f \ (a ++ b) = \text{map } f \ a ++ \text{map } f \ b$$

## More Examples

*concat* ::  $\forall a. [[a]] \rightarrow [a]$

## What's the parametricity theorem?

## Example (Answer)

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

# Higher Order Functions

*filter* ::  $\forall a. (a \rightarrow \text{Bool}) \rightarrow [a] \rightarrow [a]$

## What's the parametricity theorem?

## Example (Answer)

$$\text{filter } p \ (\text{map } f \ \text{ls}) = \text{map } f \ (\text{filter} \ (p \circ f) \ \text{ls})$$

## Parametricity Theorems

Follow a similar structure. In fact it can be mechanically derived, using the *relational parametricity* framework invented by John C. Reynolds, and popularised by Wadler in the famous paper, “Theorems for Free!”<sup>1</sup>.

Upshot: We can ask lambdabot on the Haskell IRC channel for these theorems.

<sup>1</sup><https://people.mpi-sws.org/~dreyer/tor/papers/wadler.pdf>

# COMP3141

Software System Design and Implementation

## More on the Curry Howard Isomorphism

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Term 2 2019

## What is Intuitionistic Logic?

- Classical logic is the logic that most people know about.
- Intuitionistic logic does not contain the axiom of excluded middle  $p \vee \neg p$  or equivalently  $\neg\neg p \rightarrow p$ .
- In classical logic more can be proven but less can be expressed.
- Intuitionistic proof of an existence statement gives a witness for the statement.

## Example of Existence in the Classical Sense

- Let  $\mathbb{Q}$  be the set of rational numbers and  $\mathbb{I}$  be the set of irrational numbers.
- Consider the statement  $\exists x, y. (x \in \mathbb{I}) \wedge (y \in \mathbb{I}) \wedge (x^y \in \mathbb{Q})$ .
- Proof:
  - Consider the number  $\sqrt{2}^{\sqrt{2}}$ .
  - ① If  $\sqrt{2}^{\sqrt{2}} \in \mathbb{Q}$ 
    - 
    -
  - ② Otherwise if  $\sqrt{2}^{\sqrt{2}} \in \mathbb{I}$ 
    -

## Example of Existence in the Classical Sense

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- Proof:
  - Consider the number  $\sqrt{2}^{\sqrt{2}}$ .
    - ① If  $\sqrt{2}^{\sqrt{2}} \in \mathbb{Q}$ 
      - Pick  $x = \sqrt{2}$  and  $y = \sqrt{2}$
      - Then  $x^y = \sqrt{2}^{\sqrt{2}}$  so  $x^y \in \mathbb{Q}$
    - ② Otherwise if  $\sqrt{2}^{\sqrt{2}} \in \mathbb{I}$ 
      - 
      -

## Example of Existence in the Classical Sense

- Let  $\mathbb{Q}$  be the set of rational numbers and  $\mathbb{I}$  be the set of irrational numbers.
- Consider the statement  $\exists x, y. (x \in \mathbb{I}) \wedge (y \in \mathbb{I}) \wedge (x^y \in \mathbb{Q})$ .
- Proof:
  - Consider the number  $\sqrt{2}^{\sqrt{2}}$ .
    - ① If  $\sqrt{2}^{\sqrt{2}} \in \mathbb{Q}$ 
      - Pick  $x = \sqrt{2}$  and  $y = \sqrt{2}$
      - Then  $x^y = \sqrt{2}^{\sqrt{2}}$  so  $x^y \in \mathbb{Q}$
    - ② Otherwise if  $\sqrt{2}^{\sqrt{2}} \in \mathbb{I}$ 
      - Pick  $x = \sqrt{2}^{\sqrt{2}}$  and  $y = \sqrt{2}$
      - Then  $x^y = (\sqrt{2}^{\sqrt{2}})^{\sqrt{2}} = 2$  so  $x^y \in \mathbb{Q}$

## Recall: The Curry-Howard Isomorphism

This correspondence goes by many names, but is usually attributed to Haskell Curry and William Howard.

It is a *very deep* result:

Logic	Programming
Propositions	Types
Proofs	Programs
Proof Simplification	Evaluation

It turns out, no matter what logic you want to define, there is always a corresponding  $\lambda$ -calculus, and vice versa.

Constructive Logic Classical Logic Modal Logic Linear Logic Separation Logic	Typed $\lambda$ -Calculus Continuations Monads Linear Types, Session Types Region Types
--	---

# Translating

We can translate logical connectives to types and back:

Conjunction ( $\wedge$ )	Tuples
Disjunction ( $\vee$ )	Either
Implication	Functions
True	()
False	Void

We can also translate our *equational reasoning* on programs into  
*proof simplification* on proofs!

## Examples

```
prop_or_false :: a -> (Either a void)
prop_or_false a = Left a
```

```
prop_or_true :: a -> (Either a ())
prop_or_true a = Right ()
```

```
prop_and_true :: a -> (a, ())
prop_and_true a = (a, ())
```

```
prop_double_neg_intro :: a -> (a -> void) -> void
prop_double_neg_intro a f = f a
```

```
prop_triple_neg_elim :: 
  (((a -> void) -> void) -> void) -> a -> void
prop_triple_neg_elim f a = f (\g -> g a)
```