# Proof assistants, dependent types, and modeling...?

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### Outline

### Types

### Dependent types

### Another example of dependent type

### Proofs

Modeling?

# Python

- I want to try to quickly something called *dependent types* which enable *proofs* in the context of computer code.
- The language we have used in this course python is dynamically typed:

Python is called a dynamically typed language because you do not need to declare the type of a variable when you create it; the type is determined automatically based on the value assigned to it.

# A python example

So for example we can write import numpy as np def f(a): return a + np.array([1,1,1])

without first declaring that a is an np.array of length 3. We just get a *runtime error* if a isn't of the correct form.

we get

f([1,0,-1])

array([2, 1, 0])

### continued

```
import numpy as np
def f(a):
    return a + np.array([1,1,1])
```

```
try:
    f([1,0,-1,0])
except:
    print("runtime error...")
```

runtime error...

# Statically typed language

- In contrast, in a statically typed language you have to be more explicit about things.
- Because my plan for this talk is ultimately to describe a little bit about the Lean language/proof-assistant, I'm going to discuss typing for Lean, but until I discuss dependent types, my remarks mostly describe typing for any language in the ML family (Haskell, OCaml, ...).

### An add function

- Let's define a function add that that adds 2 lists of natural numbers: we view the arguments as "vectors" and we want the function to add these vectors.
- ▶ The intent is that add [1, 1, 1] [1, -2, 1] should return something like [2, -1, 2]. In this case, the type system will only permit you to call the function add with two arguments, both required to be lists of natural numbers.
- So e.g. add ["a"] [1] should fail, but not with a runtime error – the language "knows" this invocation is prohibited because it can infer the type of ["a"] as List String instead of List N.

# Error handling via Option

```
But we have to worry about add [1,1] [1,1,1].
```

- One way to deal with this is to have a type for error handling. Here if a is a type, then Option a is the type which can have values either none or some a.
- Our function can return a Option value. The signature of our function will be

```
def add (a: List N) (b: List N)
  : Option (List N)
```

So an invocation of add can either return none or it can return some [  $\ldots$  ].

### The add function

```
def add (a: List N) (b: List N) :
  Option (List ℕ) :=
  if a.length == b.length then
    match a,b with
    [],_ => some []
    | ,[] => some []
    (c::cs), (d::ds) => do
      let rest ← add cs ds
      pure $ (c+d)::rest -- this returns
                           -- a `some`-value.
  else
    none
```

# add function results

So for example

add [1,2] [3,4]

evaluates to some [4,6] while

while

add [1,2,3] [4,5]

evaluates to none.

The main drawback with this approach is that after using it, one is then committed to carrying around values of the Option type

# A dependent type

- But Lean and other dependently typed languages such as Idris, Agda, ... – offers us something more.
- We can encode the statement "a and b have the same length" using a piece of data. We view this data as a *proof*, or as *evidence* of the equality.
- If x y : N then x = y is a type; more precisely, x = y is a Proposition in Lean.
- In contrast, x == y is really a boolean valued procedure, with signature something like

def (==) {a : Type} (x y : a) : Boolean

### equality types, continued

**For example**, Lean knows statements like

theorem eq\_succ { x y : N } :  $x = y \rightarrow (x+1) = (y+1)$ 

which we read as "if x and y are equal, then so are x+1 and v+1". (more precisely: it is easy to prove such statements using Lean)

# Type-safe add

- One way to use this equality type is to require such evidence as an argument to a function
- For example, we can require the user provide an equality proof to *invoke* our addition function.
- Here is possible type-signature for such a function

**def** add safe (a:List N) (b:List N) (p:a.length = b.length) : List N



Thus, one can make a call

#### add\_safe 11 12 p

where 11 12 : List N are lists of natural numbers, and where p : 11.length = 12.length is a proof that the lists have the same length.

Type-safe add (implemented)

here is the code

```
def add_safe (a:List N) (b:List N)
  (p:a.length = b.length) : List N :=
  match a,b with
  | [],[] => []
  | z::zs, w::ws => by
  have h : zs.length = ws.length := by
    repeat rw [List.length_cons] at p
    linarith
  exact (z+w)::add safe zs ws h
```

note that we needed to construct the proof h from the hypothesis p in order to be able to recursively invoke the function add\_safe on the shorter lists zs and ws. Type-safe add, continued



#### add\_safe [ 1,2,3] [1,2,4] rfl

evaluates to [2,4,7]. Here rfl is a proof that
[1,2,3].length = [1,2,4].length - this proof
amounts to the "reflexive law of equality".

in contrast

add\_safe [ 1,2,3] [1,2,4,5] rfl

doesn't type-check in Lean.

### Vectors

- a basic example of a dependent type is that of a vector
- ▶ the idea is that the type itself indicates how many entries the vector has. This is like saying that  $\mathbb{R}^3$  is a type
- of course, you can make a type for "3-tuples of floats" in more-or-less any typed language. But dependently typed languages let you make a type for "n-tuples of floats" where n is a variable natural number.

### vectors continued

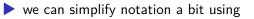
here is a definition of a vector (this isn't actually the definition used in Lean, which is more complicated for reasons that aren't really relevant to our discussion).

```
inductive vect : Type → N → Type where
| vnil : vect a 0
| vcons (x:a) (v:vect a n)
      : vect a (Nat.succ n)
```

- vect is an inductive type. There are two constructors: vnil is the "empty vector" (of length 0) and vcons constructs a vector of length n from an element and a vector of length n-1
- thus we can create a vector of length two of natural numbers

```
vect.vcons 1 (vect.vcons 2 vect.vnil)
```

```
which "is" the vector [1,2]
```



infixr:67 " ::: " => vect.vcons

now our vector representing [1, 2] above can be entered as

1 ::: 2 ::: vnil

### vectors as dependent type

```
vect a n is a dependent type
```

the type vect a n has a type parameter - in this case, a, which is an arbitrary type. But this doesn't make it a dependent type. E.g. this is essentially the same as the Option or List type constructors we have seen before, and which many non-dependently typed languages have. e.g. the definition of Option is as follows. The type doesn't depend on a value

```
inductive Option (a:Type) where
 none : Option a -- no value
 some (val:a) : Option a
```



what makes vect a n dependent is the value parameter n, a natural number

## code for adding our vectors

rather than giving our add\_safe function a proof that its arguments are equal-length lists, we can instead define an add\_vect function with signature

thus add\_vect will only accept as arguments vectors of the same length

### code for adding our vectors

The code is actually simpler than that of our earlier add\_safe:

### adding some vectors

### add\_vect (1 :::: 2 :::: 3 :::: vect.vnil) (1 :::: 1 :::: 1 :::: vect.vnil)

evaluates to 2 ::: 3 ::: 4 ::: vect.vnil

Proving statements about constructions

List in lean is another type constructor

inductive List (a:Type) where
| nil : List a -- empty list
| cons (x:a) (xs:List a) : List a

where we define notation [] for nil and x :: xs for cons x xs.

thus e.g.

1 :: 2 :: 3 :: nil

is the list [1, 2, 3]

the main difference between List and Vector of course is that Vector s have a fixed length, while List s don't

## appending lists

Here is some Lean code that appends two lists.

def append {a:Type} (xs ys : List a)
 : List a :=
 match xs with
 [] => ys
 | z :: zs => z :: append zs ys

append ["a", "b", "c"] ["d", "e"]

evaluates to ["a", "b", "c", "d", "e"]

Now, we are going to prove a property about this append function: namely, that the length of the appended lists is the sum of their lengths.

# the proof

here is the proof in Lean

```
theorem append_length {a:Type}
   (xs ys : List a)
   : (append xs ys).length =
        xs.length + ys.length := by
   induction xs with
   | nil => simp [append]
   | cons z zs ih =>
        simp [append, ih]
        linarith
```

you can view this theorem append\_length as a function of xs and ys, whose value is the indicated equality Proposition.

# the proof continued

here is the proof in Lean

```
theorem append_length {a:Type}
   (xs ys : List a)
   : (append xs ys).length =
        xs.length + ys.length := by
induction xs with
   | nil => simp [append]
   | cons z zs ih =>
        simp [append, ih]
        linarith
```

The proof is by induction on the length of the first list.

In the base case where the first list is empty, the proof boils down to the observation that append [] ys is equal to ys. We are able to produce the pf using the *simplifier tactic* simp.

# proof continued 2

here is the proof in Lean

```
theorem append_length {a:Type}
  (xs ys : List a)
  : (append xs ys).length =
      xs.length + ys.length := by
induction xs with
  | nil => simp [append]
  | cons z zs ih =>
      simp [append, ih]
      linarith
```

when the first list is non-empty, it must the form z::zs and we then have the inductive hypotheses that append zs ys has length equal to zs.length + ys.length. Using this, the required result is again provided by simp.

### Relevance to modeling?

Some thoughts, remarks, and questions:

- would more type-safe linear algebra functions be useful in the setting of math modeling?
  - advantages: types help prevent certain types of errors, and proofs provide assurance of correctness
  - disadvantages: more complex to create code
- one can imagine proofs of properties of results obtained from modeling software; how useful would this be?
- I'm aware of a fair amount of recent formalization activity in pure mathematics, but I know less about its adoption in math modeling settings
- having a machine-usable language for mathematical proofs is a pre-requisite for doing machine-learning *about* mathematical statements