

# **Diffeological spaces as a model for differentiable programs**

A tutorial

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these slides available at [philipsaville.co.uk](http://philipsaville.co.uk)

- (1) What questions does **denotational semantics** study?
- (2) Why are **cartesian closed categories** so important?
- (3) Where do **diffeological spaces** come in?

# What does programming language theory study?

We want programs that are:

efficient, fast, and correct

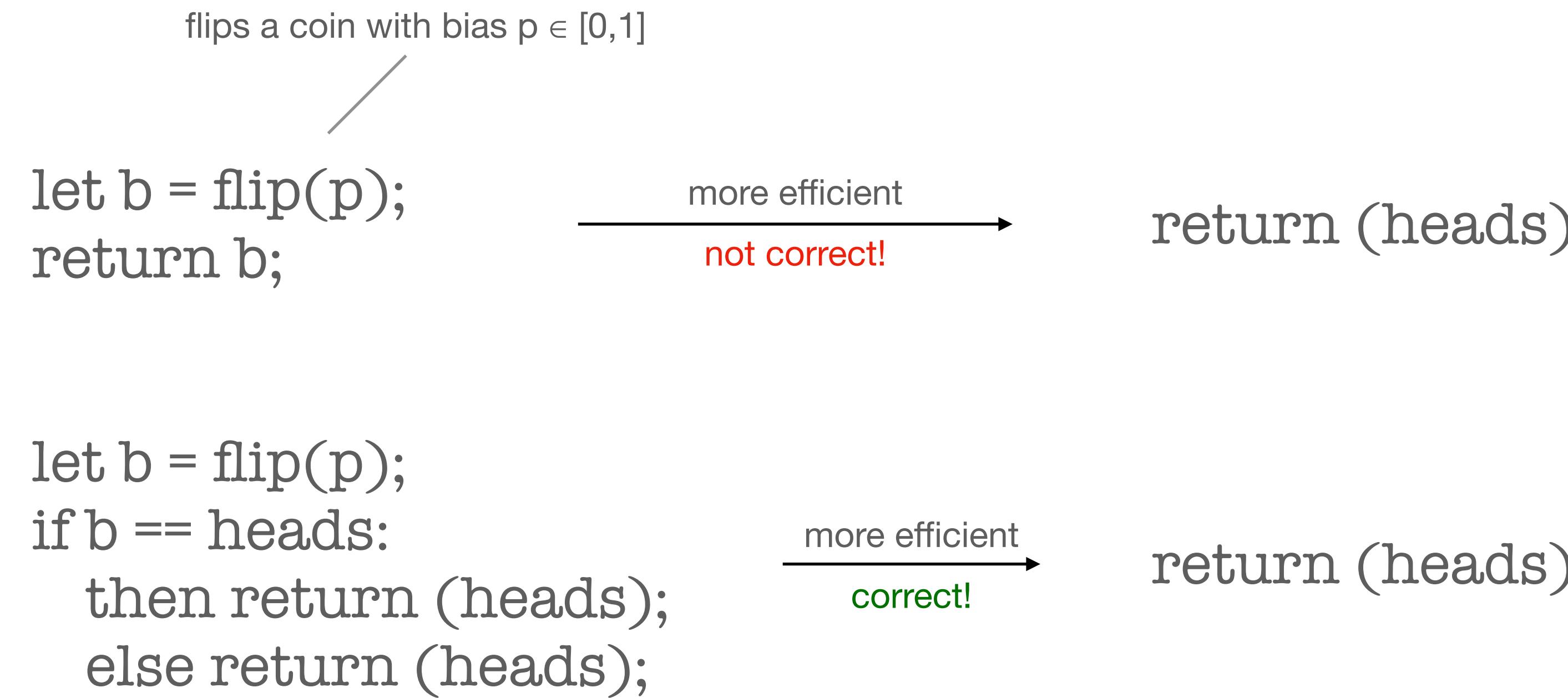
We ask:

- (1) When are programs interchangeable?
- (2) How should we think about programs?

gets interesting when programs have effects  
= interaction with the world

# When are programs interchangeable?

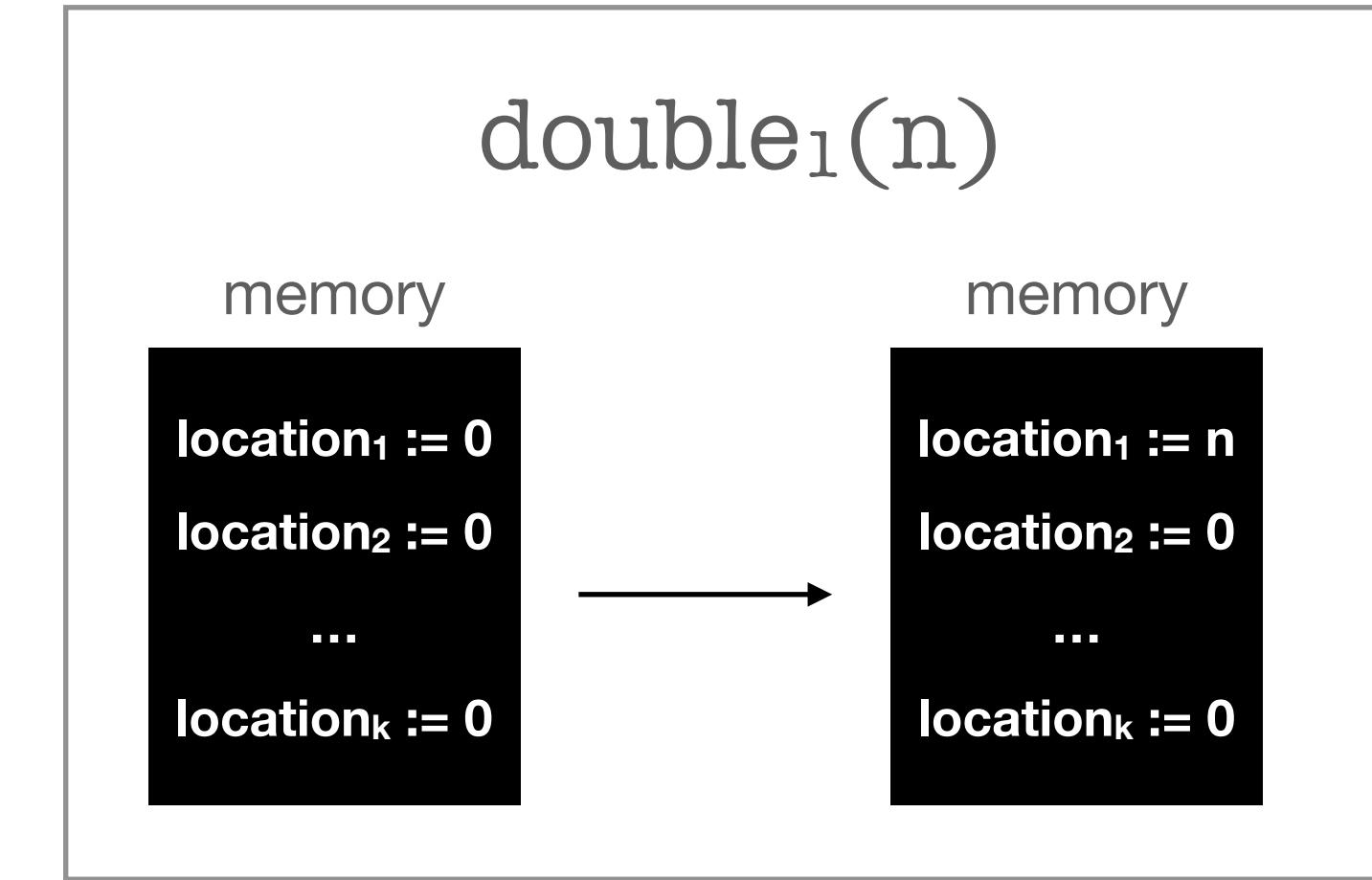
example 1



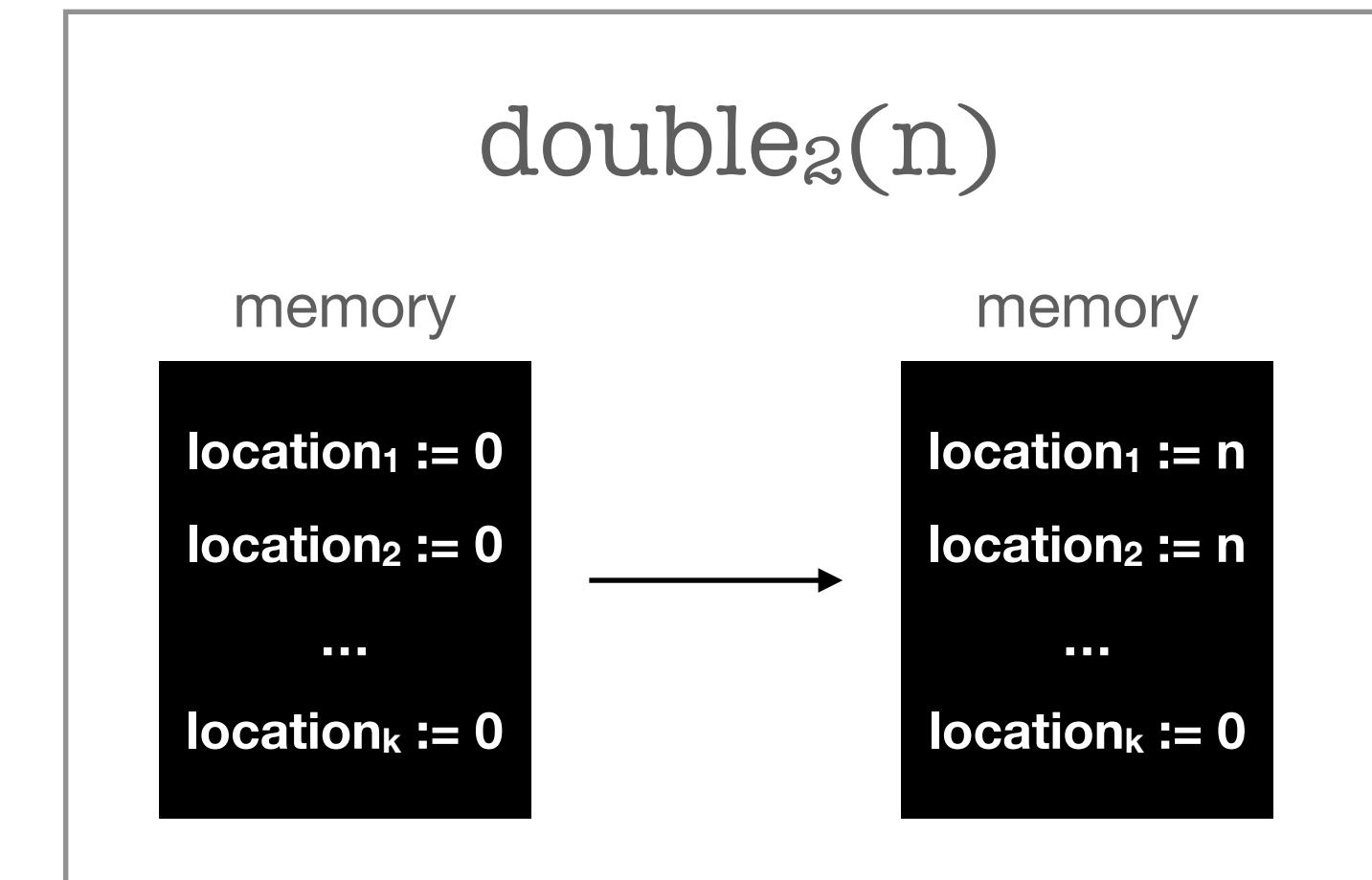
# When are programs interchangeable?

example 2

```
fun double1(n):
    set_memory location1 := n;
    return (
        get_memory (location1)
        + get_memory (location1)
    );
```



```
fun double2(n):
    set_memory location1 := n
    set_memory location2 := n
    return (
        get_memory (location1)
        + get_memory (location2)
    );
```

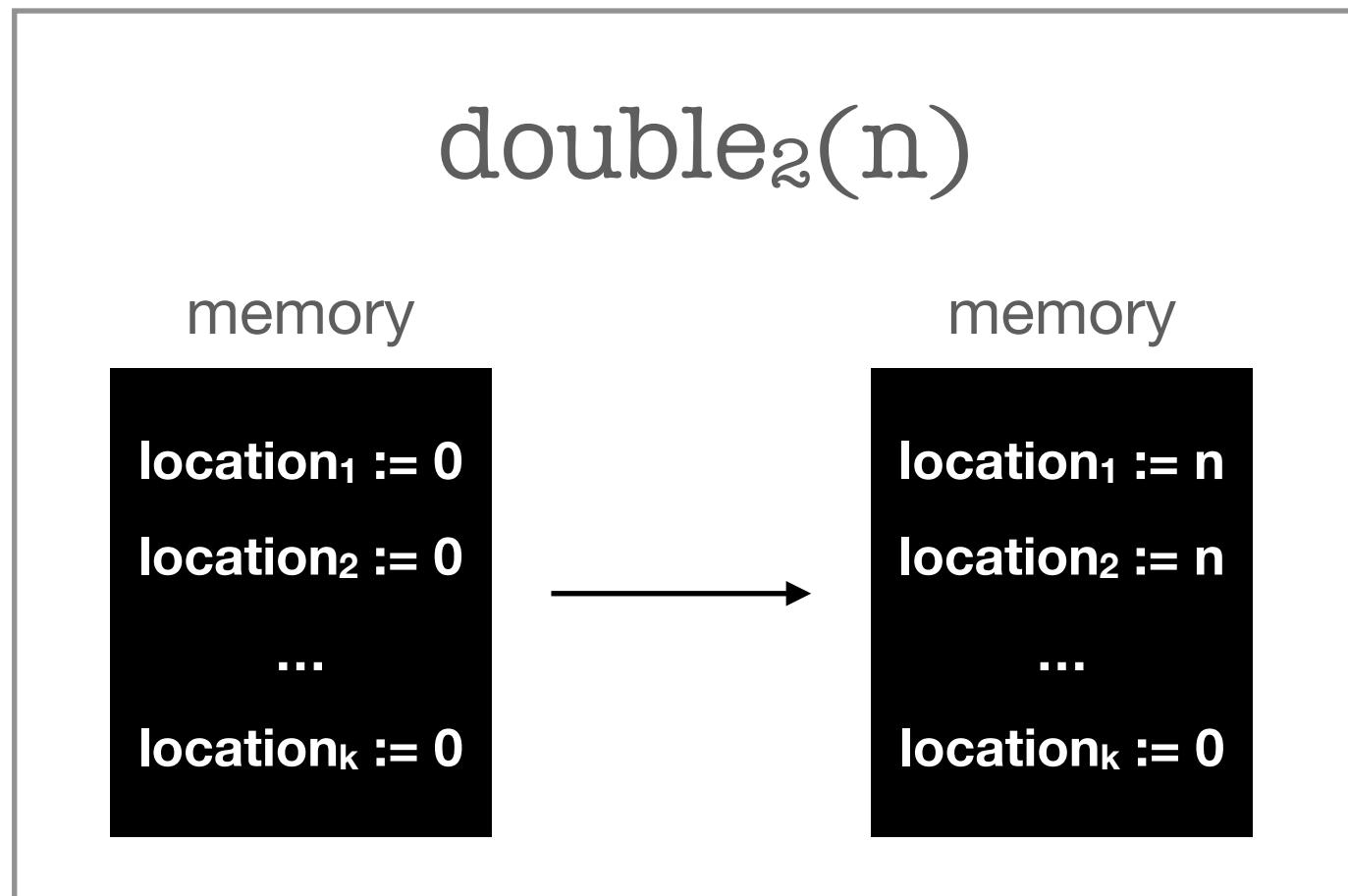
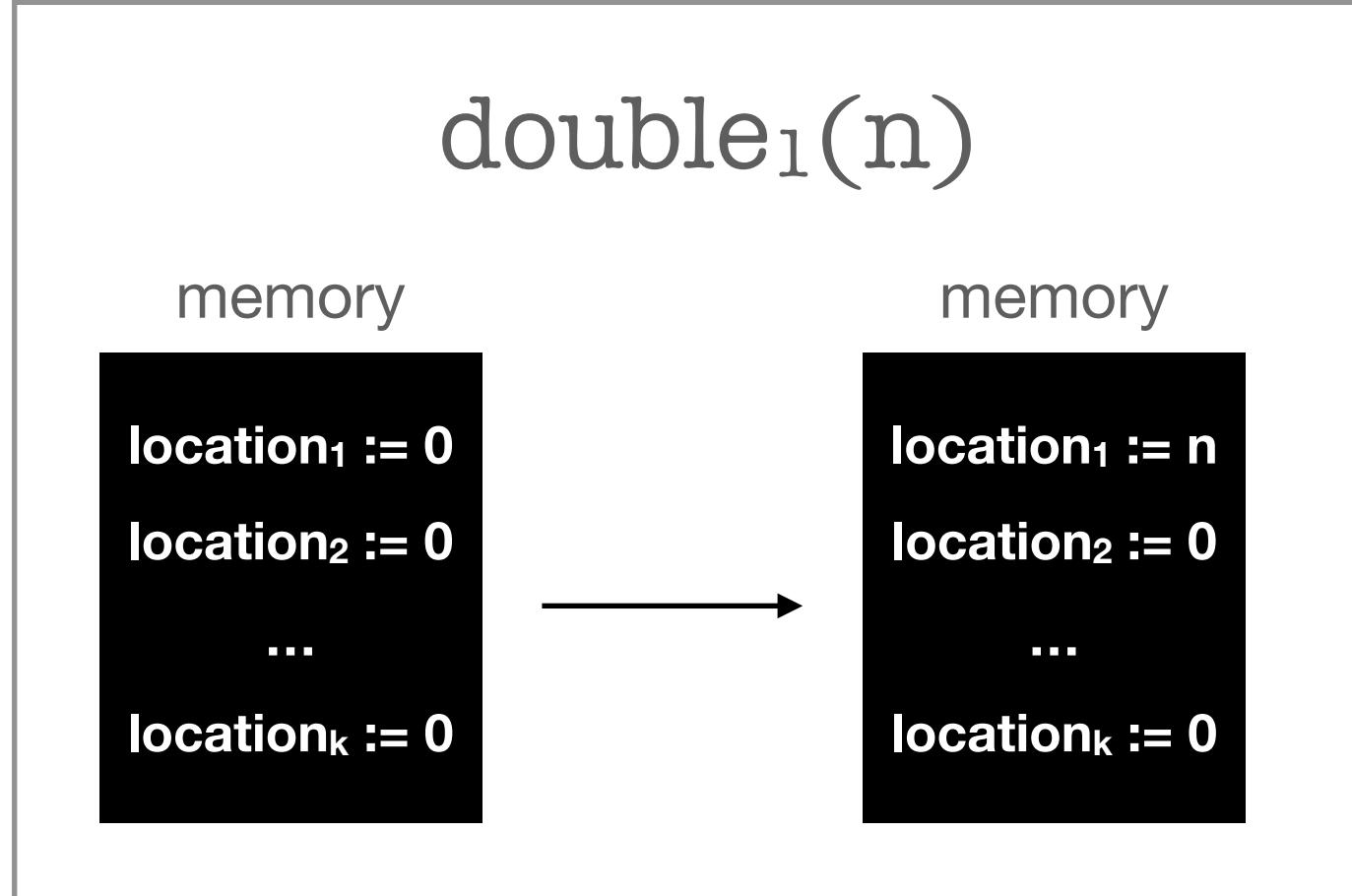


# When are programs interchangeable?

example 2

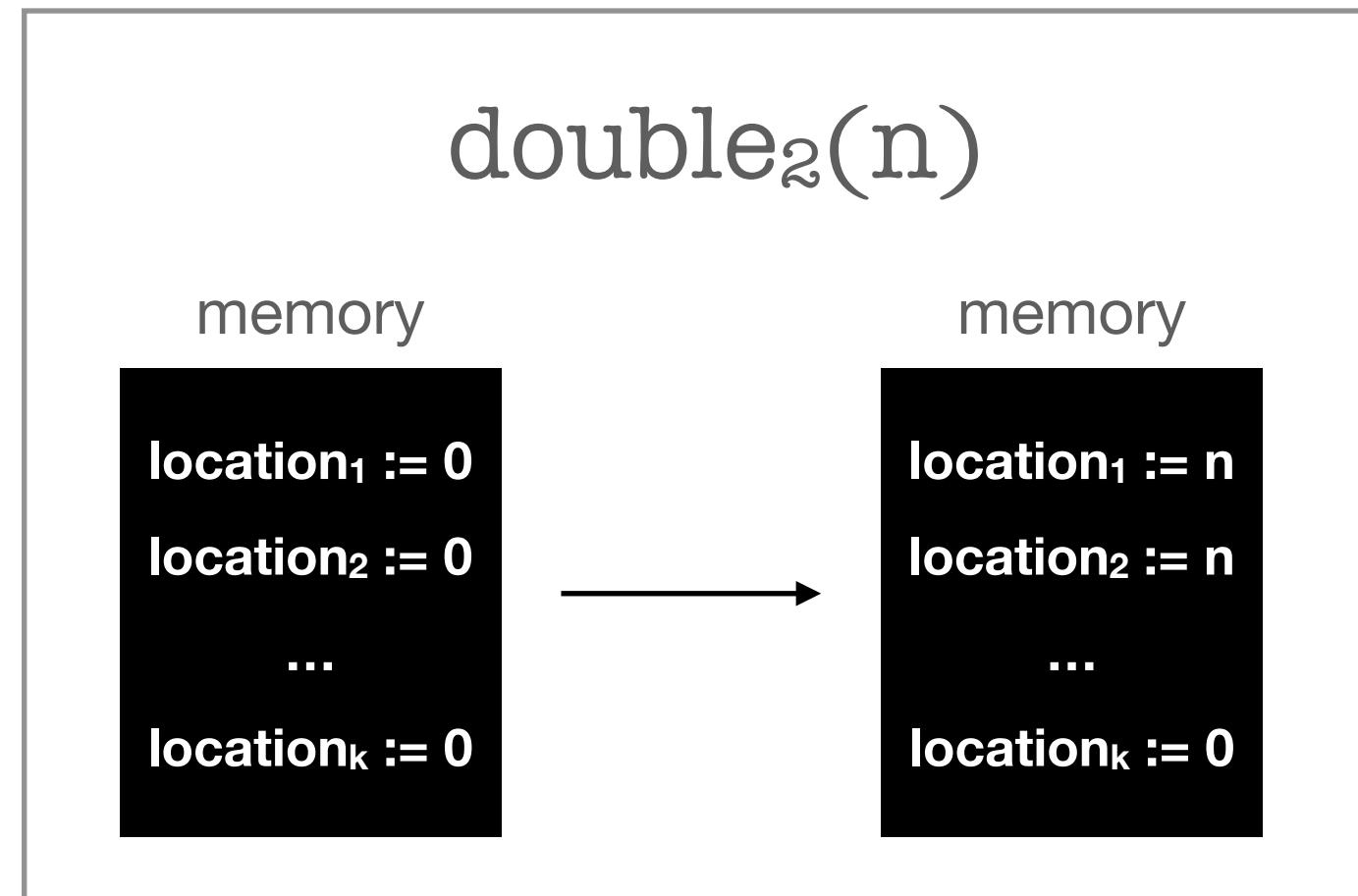
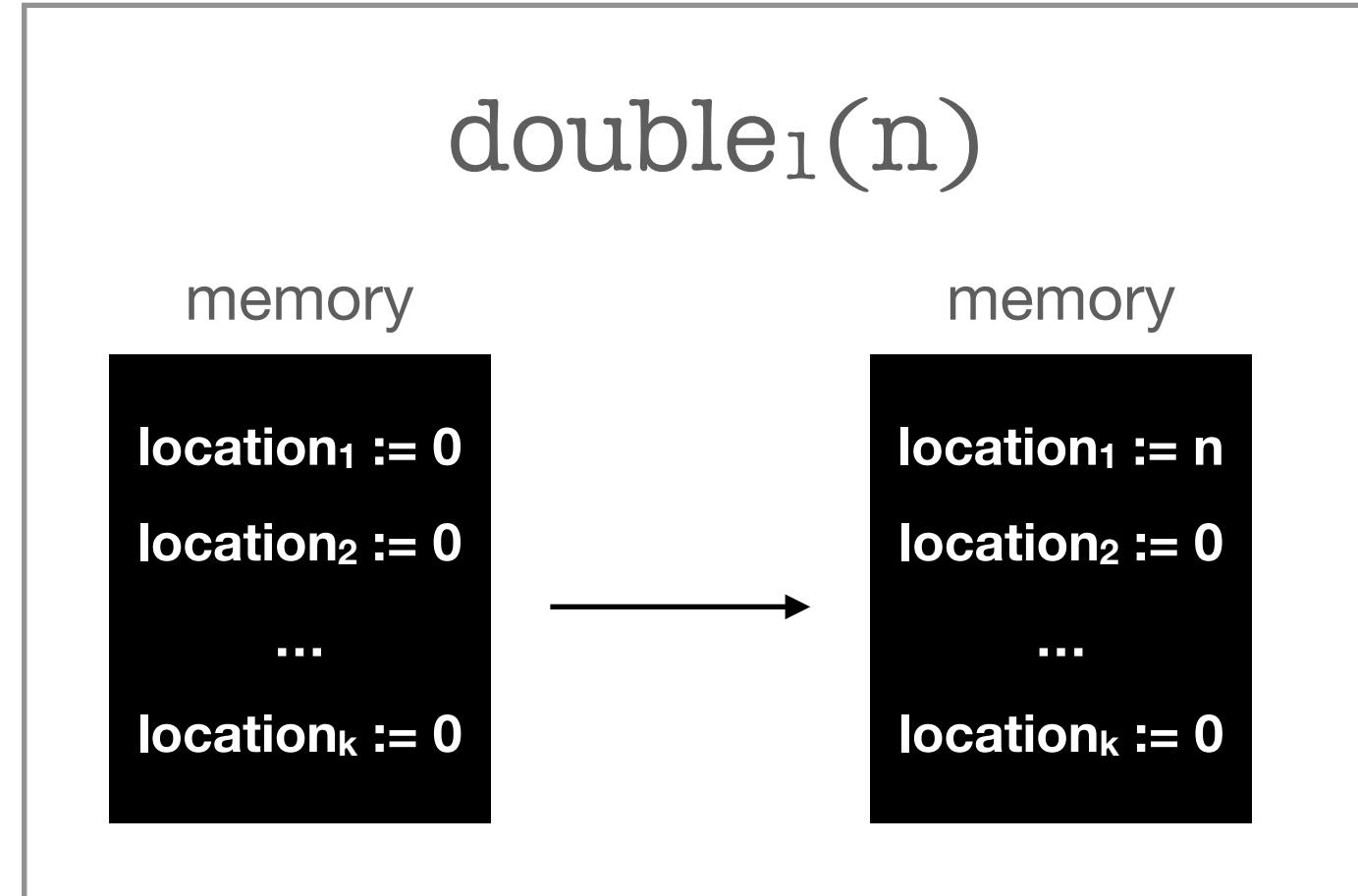
equal as **functions** but not as **programs**!

↗ we can observe a difference in behaviour



# When are programs interchangeable?

example 2



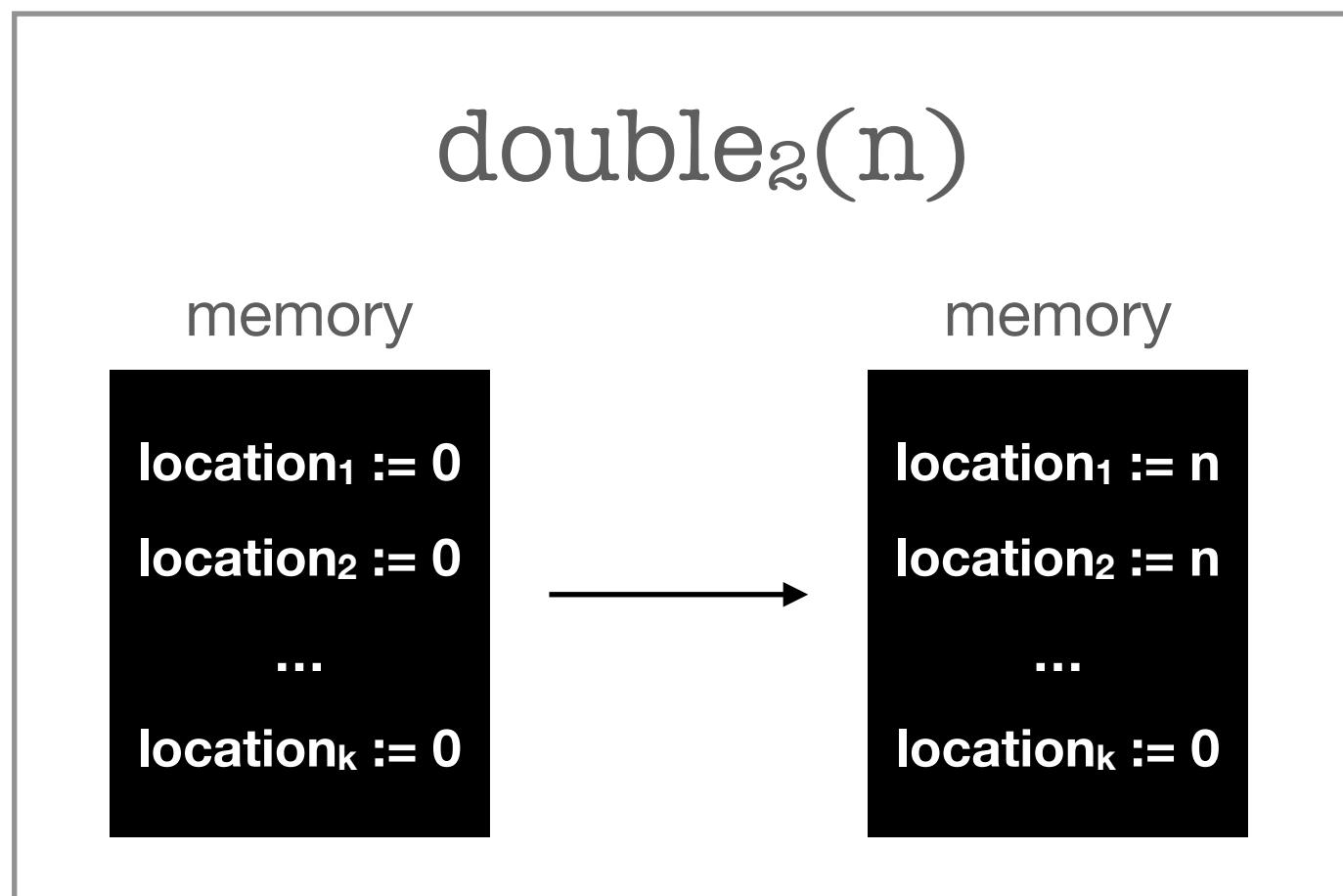
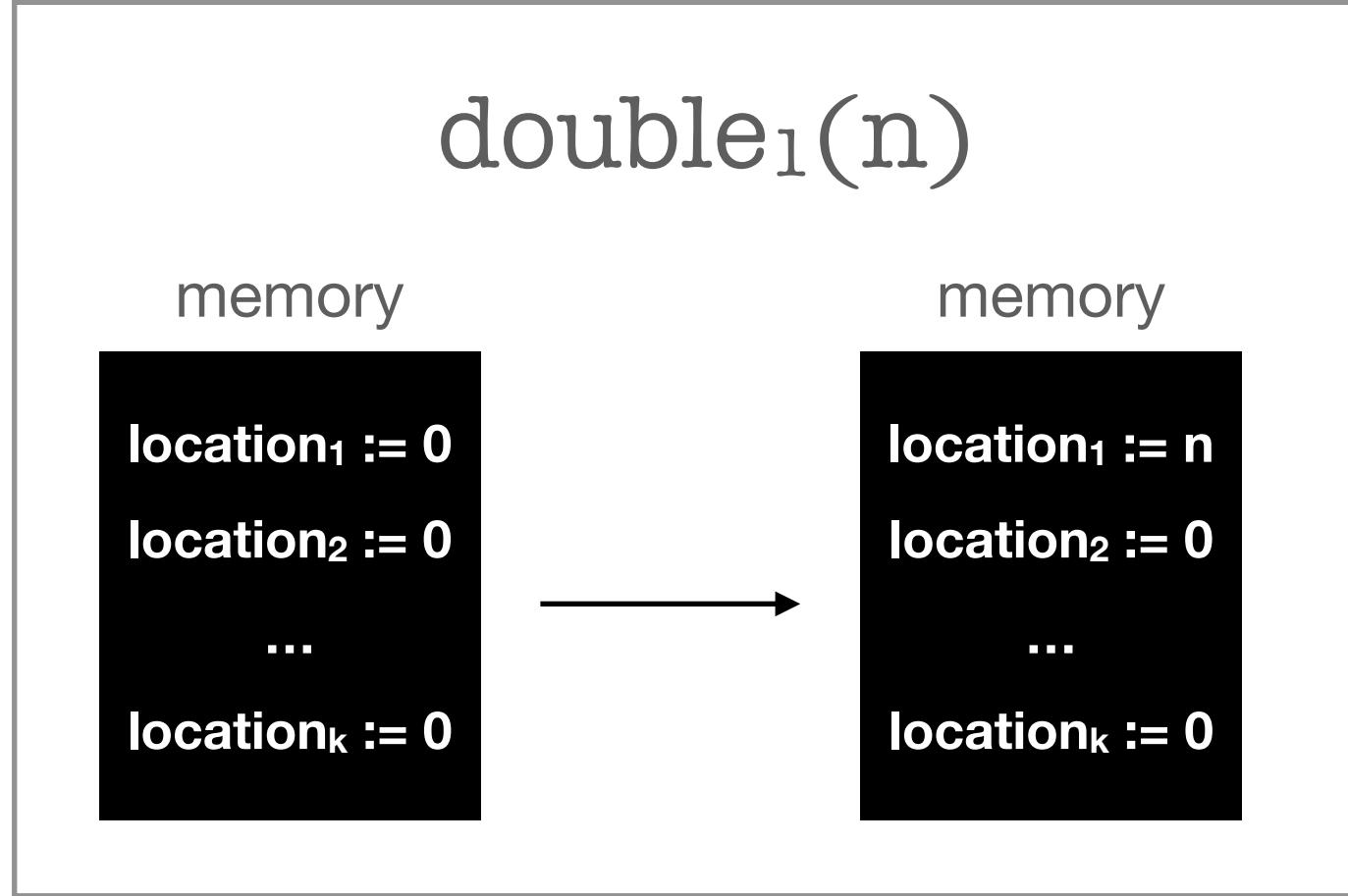
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$$\forall n . \text{double}_1(n) = \text{double}_2(n)$$

# When are programs interchangeable?

example 2



equal as **functions** but not as **programs**!

→ we can observe a difference in behaviour

$$\forall n . \text{double}_1(n) = \text{double}_2(n)$$

but can distinguish them by looking at memory:

```
doublei(2);  
let n = get_memory location2;  
if n > 0:  
    then return (false);  
else return (true);
```

A flowchart illustrating the execution of the code. It starts with "double<sub>i</sub>(2);". An arrow leads to "let n = get\_memory location<sub>2</sub>;". From there, two paths emerge: one for "i = 1" leading to "return (true);", and one for "i = 2" leading to "return (false);".

# When are programs **interchangeable**?

programs  $P$  and  $Q$  are **observationally equivalent**  
if there's no way to observe a difference in behaviour

any program  $\mathcal{C}[P]$  containing  $P$  gives a result  
iff  $\mathcal{C}[Q]$  gives the same result

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observational equivalence

generally harder than

function equality

depends on the

language's features

depends on how  
programs run

# Observational equivalence in the real world

## how do you prove you're not Banksy?

**A town councillor has resigned, blaming people who falsely accused him of being the world famous artist Banksy.**

Pembroke Dock councillor William Gannon said the "quite ridiculous" claims were made on several social media pages.

In his resignation letter he claimed this was "undermining my ability to do the work" of a councillor.

Mr Gannon has since made an "I am not Banksy" badge to avoid any confusion and said he would now be returning to his former role of community artist.

He said the allegations meant people were "asking me to prove who I am not and that's almost impossible to do".

<https://www.bbc.co.uk/news/uk-wales-61552865>



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if there's no way to tell them apart, they must be the same!

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# How should we think about programs?

```
fun add(x, y):  
    return (x + y)
```

a function  $\mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$

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```
fun divide(x, y):  
    return (x / y)
```

a function  $\mathbb{Z} \times \mathbb{Z}_{\neq 0} \rightarrow \mathbb{Q}$

a function  $\mathbb{Z} \times \mathbb{Z} \rightarrow \mathbb{Q} + \{\text{fail}\}$

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```
fun print_and_return(x):  
    print "hello";  
    return x;
```

a function  $\mathbb{N} \rightarrow \{a, b, \dots, z\}^* \times \mathbb{N}$   
 $x \mapsto (\text{hello}, x)$

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```
let b = flip(p);  
return b;
```

a probability distribution on  
 $\{\text{heads}, \text{tails}\}$

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let b = flip(p);  
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a probability distribution on  
{heads, tails}

```
normalise(  
    let x = sample(bernoulli(0.8));  
    let r = (if x then 10 else 3);  
    observe 0.45 from exponential(r)  
    return(x)  
)
```

some measurable function (??)

# What does programming language theory study?

We want programs that are:  
**efficient, fast, and correct**

We ask:

- (1) When are programs interchangeable?
- (2) How should we think about programs?

some kind  
of function?

need something beyond set-theoretic  
functions to model richer features!



uses ideas from:  
- topology  
- logic  
- order theory

**observational equivalence**  
generally harder than  
function equality  
depends on the  
language's features

depends on how  
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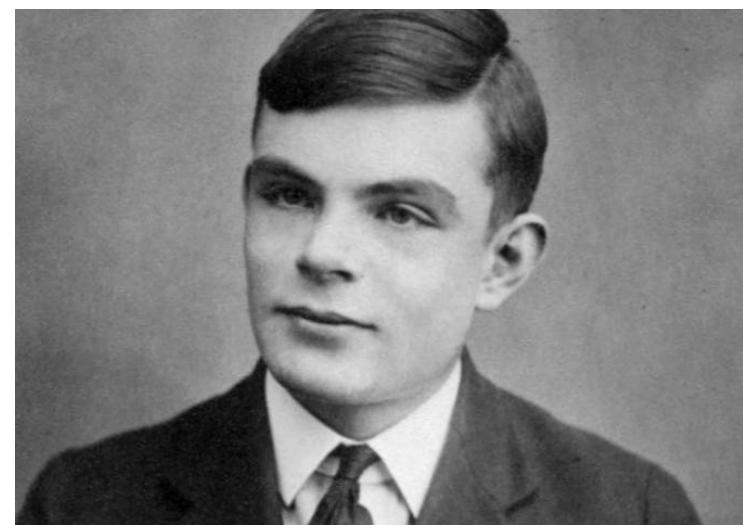
The denotational semantics perspective:

- (1) Assign every program  $P$  a meaning  $\llbracket P \rrbracket$
- (2) Reason about equality of programs via their meaning
- (3) The semantic model tells you what programs ‘really are’

# Coming up next

1. Introduce an idealised functional programming language
  2. Explain its semantic interpretation in CCCs
- 
3. Introduce differentiable programming
  4. Explain the interpretation in Diff

# What is a program?



something modelled by a Turing machine

memory you can read  
to & write from



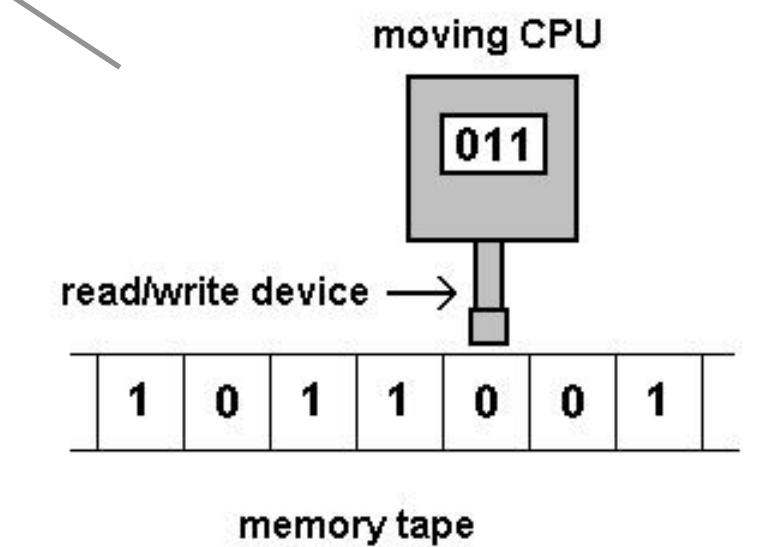
a kind of function

OCaml, Haskell,  
Standard ML,...

fib 0 = 0  
fib 1 = 1  
fib n = fib (n-1) + fib (n-2)

Java, C, C++, ...

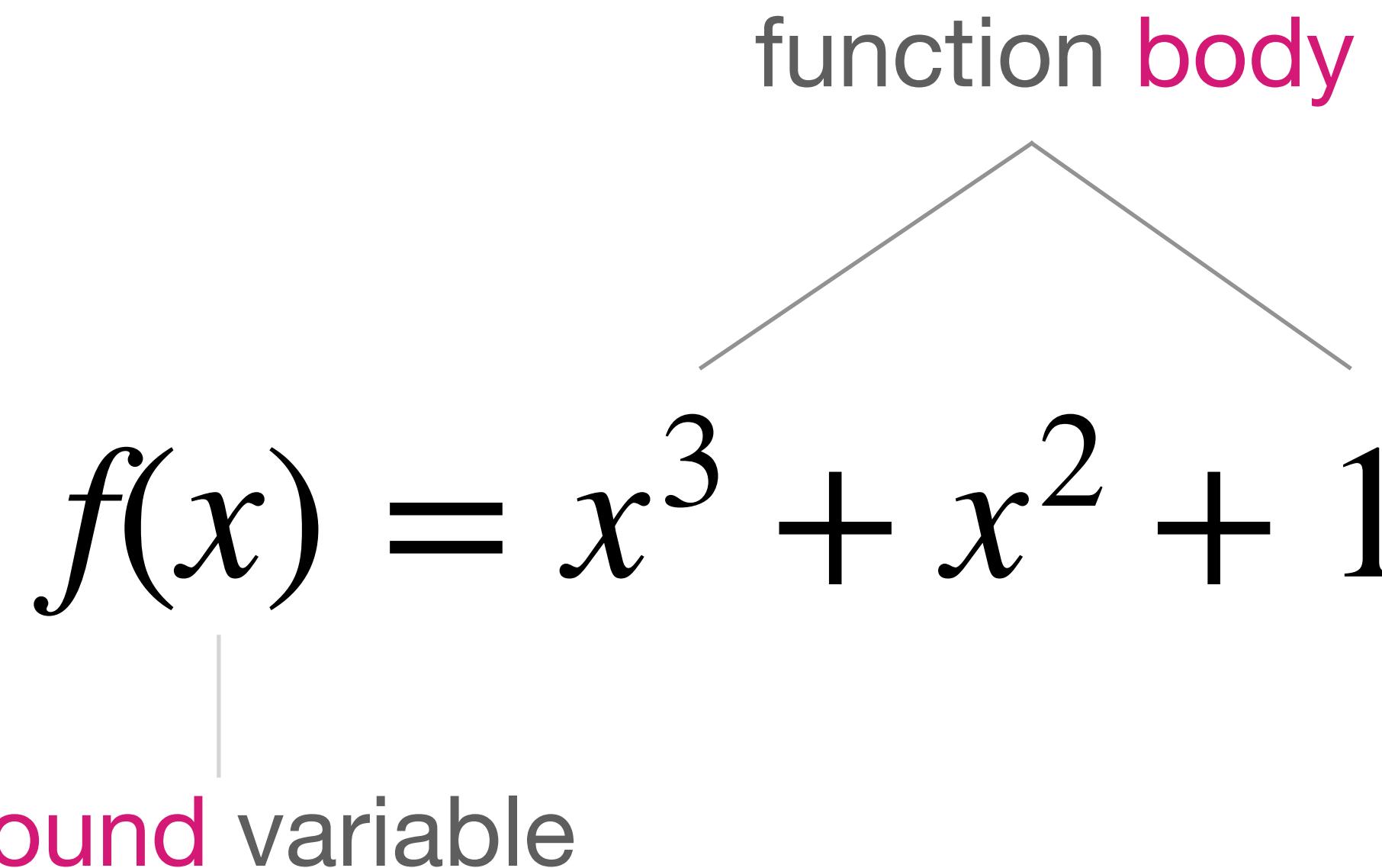
```
n1 = 0
n2 = 1
steps_taken = 0
while (steps_taken < 100) {
    fib = n1 + n2
    n1 = n2
    n2 = fib
    steps_taken = steps_taken + 1
}
```



So a **functional programming language** lets you

- form functions
- evaluate functions at arguments

# How do we define functions?



may not use  $x$ , eg  $f(x) = 3$

may contain free variables, eg  
 $f(x) = 3y + x$

A **functional programming language** lets you

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the  $x$  matters: if

$$g(y) = 3y^3 + y^2 + 1$$

$$h(y) = 3x^3 + x^2 + 1$$

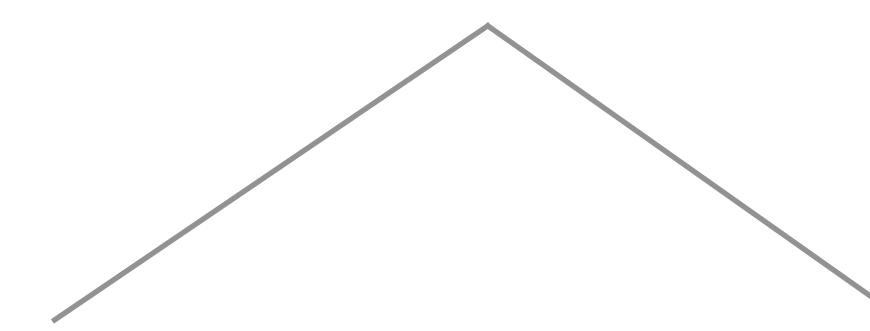
then  $f = g$  but  $h$  is a constant function

every other  
variable is **free**

# How do we define functions?



function **body**



may not use  $x$ , eg  $f(x) = 3$

may contain free variables, eg  
 $f(x) = 3y + x$

$$f(x) = x^3 + x^2 + 1$$

**bound variable**

the  $x$  matters: if

$$g(y) = 3y^3 + y^2 + 1$$

$$h(y) = 3x^3 + x^2 + 1$$

then  $f = g$  but  $h$  is a constant function

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variable is **free**

evaluating = **substituting** for bound variable

$$f(3) = (x^3 + x^2 + 1)[x \mapsto 3]$$

$$= 3^3 + 3^2 + 1$$

# How do we define functions?



function **body**

may not use  $x$ , eg  $f(x) = 3$

may contain free variables, eg  
 $f(x) = 3y + x$

$$f(x) = x^3 + x^2 + 1 \text{ in } \mathbb{R} \text{ whenever } x \in \mathbb{R}$$

**bound variable**  $x \in \mathbb{R}$

evaluating = **substituting** for bound variable

the  $x$  matters: if

$$\begin{aligned}g(y) &= 3y^3 + y^2 + 1 \\h(y) &= 3x^3 + x^2 + 1\end{aligned}$$

then  $f = g$  but  $h$  is a constant function

every other variable is **free**

$$\begin{aligned}f(3) &= (x^3 + x^2 + 1)[x \mapsto 3] \\3 \in \mathbb{R} \\&= 3^3 + 3^2 + 1\end{aligned}$$

# How do we define functions?



$$\frac{\begin{array}{c} \text{function body} \\ x^3 + x^2 + 1 \text{ is a program} \end{array}}{(x \mapsto x^3 + x^2 + 1) \text{ is a program}}$$

|  
bound variable  
every other variable is free

$$\frac{(x \mapsto x^3 + x^2 + 1) \text{ is a program} \quad 3 \text{ is a program}}{(x \mapsto x^3 + x^2 + 1)(3) \text{ is a program}}$$

evaluating = substituting for bound variable

$$(x \mapsto x^3 + x^2 + 1)(3) \rightsquigarrow 3^3 + 3^2 + 1$$

extensionality:  $f = (x \mapsto f(x))$

$$(x \mapsto (x^3 + x^2 + 1)(x)) \rightsquigarrow (x \mapsto x^3 + x^2 + 1)$$

# How do we define functions?

the  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$



$$\frac{\begin{array}{c} \text{function body} \\ x^3 + x^2 + 1 \text{ is a program} \end{array}}{\lambda x . x^3 + x^2 + 1 \text{ is a program}}$$

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$$\frac{\begin{array}{c} \lambda x . x^3 + x^2 + 1 \text{ is a program} \\ 3 \text{ is a program} \end{array}}{(\lambda x . x^3 + x^2 + 1)(3) \text{ is a program}}$$

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# How do we define functions?

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$



function body  
 $x^3 + x^2 + 1$  is a program of type  $\mathbb{R}$

---

$$\lambda x . x^3 + x^2 + 1 \text{ is a program of type } \mathbb{R} \rightarrow \mathbb{R}$$

bound variable

every other variable is free

$$\lambda x . x^3 + x^2 + 1 \text{ is a program of type } \mathbb{R} \rightarrow \mathbb{R} \quad 3 \text{ is a program of type } \mathbb{R}$$

---

$$(\lambda x . x^3 + x^2 + 1)(3) \text{ is a program of type } \mathbb{R}$$

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# How do we define functions?

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$

function **body**

$P$  is a program of type  $B$

$x$  is a variable of type  $A$

abstraction

$$\lambda x . P \text{ is a program of type } A \rightarrow B$$

bound variable

every other  
variable is **free**

$P$  is a program of type  $A \rightarrow B$

$Q$  is a program of type  $A$

application

$$P(Q) \text{ is a program of type } B$$

evaluating = **substituting** for bound variable

$$(\lambda x . P)(Q) \rightsquigarrow_{\beta} P[x \mapsto Q]$$

extensionality:  $f = (x \mapsto f(x))$

$$P \rightsquigarrow_{\eta} \lambda x . P(x)$$



A **functional programming language** lets you

- form functions
- evaluate functions at arguments

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$x$  is a variable of type  $A$

$x$  is a program of type  $A$



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$$\lambda x . f(x) = (x \mapsto f(x))$$

$x$  is a variable  
\_\_\_\_\_  
 $x$  is a program

$P : B$  is a program  
abstraction  
\_\_\_\_\_  
 $\lambda x . P : A \rightarrow B$  is a program

$P : A \rightarrow B$  is a program       $Q : A$  is a program  
\_\_\_\_\_  
 $P(Q) : B$  is a program

application

evaluating = substituting for bound  $x$   
 $P(Q) \rightsquigarrow_{\beta} P[x \mapsto Q]$   
= running the program

extensionality  
 $P \rightsquigarrow_{\eta} \lambda x . P(x)$   
 $f = (x \mapsto f(x))$



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# How do we define functions?

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$

$x \text{ is a variable}$ $\lambda x . f(x) = (x \mapsto f(x))$	$\frac{}{x \text{ is a program}}$ $P : B \text{ is a program}$ $\lambda x . P : A \rightarrow B \text{ is a program}$	$x : A \text{ is a variable}$ $\lambda x . P : A \rightarrow B \text{ is a program}$
	$\frac{P : A \rightarrow B \text{ is a program} \quad Q : A \text{ is a program}}{P(Q) : B \text{ is a program}}$ <small>application</small>	

evaluating = **substituting** for bound  $x$   
 $P(Q) \rightsquigarrow_{\beta} P[x \mapsto Q]$   
**= running the program**

extensionality  
 $P \rightsquigarrow_{\eta} \lambda x . P(x)$   
 $f = (x \mapsto f(x))$



A **functional programming language** lets you

- form functions
- evaluate functions at arguments

$$\begin{array}{c}
 f : A \rightarrow B \qquad \qquad \qquad x : A \\
 \hline
 f(x) : B \\
 \hline
 \lambda x . f(x) : A \rightarrow B \\
 \hline
 \lambda f . \lambda x . f(x) : (A \rightarrow B) \rightarrow (A \rightarrow B)
 \end{array}$$

eval :  $(A \Rightarrow B) \times A \rightarrow B$

$$(f, x) \mapsto f(x)$$

via **currying**  $X \rightarrow (A \Rightarrow B) \cong (X \times A) \rightarrow B$

# How do we define functions?



the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$

$$\frac{x \text{ is a variable}}{x \text{ is a program}}$$

$$\frac{\begin{array}{c} P : B \text{ is a program} \\ \text{abstraction} \end{array} \quad x : A \text{ is a variable}}{\lambda x . P : A \rightarrow B \text{ is a program}}$$

$$\frac{\begin{array}{c} P : A \rightarrow B \text{ is a program} \\ Q : A \text{ is a program} \end{array}}{P(Q) : B \text{ is a program}}$$

application

evaluating = substituting for bound  $x$

$$P(Q) \rightsquigarrow_{\beta} P[x \mapsto Q]$$

= running the program

$$\frac{\begin{array}{c} P \rightsquigarrow_{\eta} \lambda x . P(x) \\ \text{extensionality} \end{array}}{f = (x \mapsto f(x))}$$

A functional programming language lets you

- form functions
- evaluate functions at arguments

$$\frac{f : A \rightarrow B \quad x : A}{f(x) : B}$$

$$\frac{}{\lambda x . f(x) : A \rightarrow B}$$

$$\frac{}{\lambda f . \lambda x . f(x) : (A \rightarrow B) \rightarrow (A \rightarrow B)}$$

$$\text{eval} : (A \Rightarrow B) \times A \rightarrow B$$

$$(f, x) \mapsto f(x)$$

$$\text{via currying } X \rightarrow (A \Rightarrow B) \cong (X \times A) \rightarrow B$$

$$\frac{\begin{array}{c} f : A \rightarrow B \quad x : A \\ g : B \rightarrow C \end{array}}{\frac{f(x) : B}{g(f(x)) : C}}$$

$$\frac{}{\lambda x . g(fx) : A \rightarrow C}$$

$$\frac{}{\lambda f . \lambda x . g(fx) : (A \rightarrow B) \rightarrow (A \rightarrow C)}$$

$$\lambda g . \lambda f . \lambda x . g(fx) : (B \rightarrow C) \rightarrow ((A \rightarrow B) \rightarrow (A \rightarrow C))$$

$$\text{comp} : (B \Rightarrow C) \times (A \Rightarrow B) \rightarrow (A \Rightarrow C)$$

$$(g, f) \mapsto g \circ f$$

# Things we can't do

the  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$

$$\frac{x \text{ is a variable}}{x \text{ is a program}}$$

$$\frac{P \text{ is a program}}{\lambda x . P \text{ is a program}} \text{ abstraction}$$

$$P \text{ is a program}$$

$$Q \text{ is a program}$$

application

$$P(Q) \text{ is a program}$$

evaluating = substituting for bound variable

$$\begin{aligned} P(Q) &\rightsquigarrow_{\beta} P[x \mapsto Q] \\ &= \text{running the program} \end{aligned}$$



A functional programming language lets you

- form functions
- evaluate functions at arguments

Note there's no restrictions on either rule!

$$\frac{f \text{ is a variable}}{f \text{ is a program}}$$

$$\frac{f(f) \text{ is a program}}{\lambda f . f(f) \text{ is a program}}$$

Looping, recursion, ...

$$\begin{aligned} (\lambda f . f(f)) (\lambda f . f(f)) &\rightsquigarrow (\lambda f . f(f)) [f \mapsto (\lambda f . ff)] \\ &= (\lambda f . f(f)) (\lambda f . f(f)) \end{aligned}$$

Encode Peano arithmetic

$$1 := (\lambda f . \lambda f . f(x))$$

$$2 := (\lambda f . \lambda f . f(fx))$$

$$\text{plus} := (\lambda m . \lambda n . \lambda f . \lambda x . mf(nfx))$$

# Adding primitives

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$



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$$\frac{}{\underline{n} : \text{nat}} \quad (n \in \mathbb{N})$$

$$\frac{}{\text{true} : \text{bool}} \quad \frac{}{\text{false} : \text{bool}}$$

$$\frac{}{\text{flip()} : \text{bool}}$$

# Adding primitives

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$



$$\frac{}{\underline{n} : \text{nat}} \quad (n \in \mathbb{N})$$

$$\frac{}{\text{true} : \text{bool}}$$

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$$\frac{}{\text{flip()} : \text{bool}}$$

What about plus, if etc?

$$\text{plus} : \mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$$

$$\text{if} : 2 \times \mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$$

$$\text{if}(i, n, m) = \begin{cases} n & \text{if } i = 0 \\ m & \text{if } i = 1 \end{cases}$$

A functional programming language lets you

- form functions
- evaluate functions at arguments

# Adding primitives

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$



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Option 1:  $\text{if}(b, n, m) : \text{nat}$  (where  $b : \text{bool}, n : \text{nat}, m : \text{nat}$ )

Option 2: add a type to model  $\mathbb{N} \times \mathbb{N}$

in general: introduce new types for new kinds of structure

# Adding product types

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$



How does  $X \times Y$  behave in Set?

$$\frac{x \in X \quad y \in Y}{(x, y) \in X \times Y} \text{ pair}$$

$$\frac{p \in A_1 \times A_2}{\pi_i(p) \in A_i} \text{ proj } (i = 1,2)$$

project out a pair

$$\pi_i(x_1, x_2) = x_i$$

extensionality: a pair is determined by its projections

$$p = (\pi_1(p), \pi_2(p))$$

A functional programming language lets you

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# Adding product types

the simply-typed  $\lambda$ -calculus

$$\lambda x . f(x) = (x \mapsto f(x))$$



How does  $X \times Y$  behave in simply-typed  $\lambda$ -calculus?

$$\frac{P_1 : A_1 \quad P_2 : A_2}{\langle P_1, P_2 \rangle : A_1 \times A_2} \text{ pair}$$

$$\frac{P : A_1 \times A_2}{\pi_i(P) : A_i} \text{ proj } (i = 1,2)$$

project out a pair

$$\pi_i \langle P_1, P_2 \rangle \rightsquigarrow_{\beta} P_i$$

extensionality: a pair is determined by its projections

$$P \rightsquigarrow_{\eta} \langle \pi_1(P), \pi_2(P) \rangle$$

A functional programming language lets you

- form functions
- evaluate functions at arguments

# The simply-typed $\lambda$ -calculus with products and primitives

= the simplest (typed) functional programming language

can also add sums / disjoint unions, lists, recursion, ....

$$\frac{x : A \quad P : B}{\lambda x . P : A \rightarrow B} \text{ abstraction}$$

$$\frac{P : A \rightarrow B \quad Q : A}{P(Q) : B} \text{ application}$$

evaluating = substituting for bound variable

$$(\lambda x . P)(Q) \rightsquigarrow_{\beta} P[x \mapsto Q]$$

= running the program

extensionality:  $f = (x \mapsto f(x))$

$$P \rightsquigarrow_{\eta} \lambda x . P(x)$$

+ any others you might want!

$$\text{true} : \text{bool}$$

$$\text{false} : \text{bool}$$

$$\text{if}(b, n, m) : \text{nat}$$

$$\text{flip()} : \text{bool}$$

$$\frac{(n \in \mathbb{N})}{\underline{n} : \text{nat}}$$

$$\text{plus} : \text{nat} \times \text{nat} \rightarrow \text{nat}$$

$\text{plus}\langle \underline{3}, \underline{2} \rangle : \text{nat}$

$\text{if}(\text{true}, \underline{3}, \underline{2}) : \text{nat}$

$$\frac{P_1 : A_1 \quad P_2 : A_2}{\langle P_1, P_2 \rangle : A_1 \times A_2} \text{ pair}$$

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extensionality: a pair is determined by its projections

$$P \rightsquigarrow_{\eta} (\pi_1(P), \pi_2(P))$$

+ a ‘unit’ type

# $\beta$ -reduction = running the program

$$(\lambda p : \text{nat} \times \text{nat} \rightarrow \text{bool} . \lambda t : \text{nat} \times \text{nat} . \text{if}(p(t), \underline{2}, \underline{3}))(\text{greater\_than})(\langle \underline{5}, \underline{6} \rangle)$$
$$\rightsquigarrow_{\beta} (\lambda t : \text{nat} \times \text{nat} . \text{if}(\text{greater\_than}(t), \underline{2}, \underline{3}))(\langle \underline{5}, \underline{6} \rangle)$$
$$\rightsquigarrow_{\beta} \text{if}(\text{greater\_than}\langle \underline{5}, \underline{6} \rangle, \underline{2}, \underline{3})$$
$$\rightsquigarrow_{\beta} \underline{3}$$

# The magic of higher-order functions

higher-order functions = functions of type  $(A \rightarrow B) \rightarrow C$

higher-order functions let you re-use code in a very efficient way

$$P : ((\text{nat} \rightarrow \text{bool}) \times (\text{nat} \rightarrow \text{nat})) \rightarrow \text{nat}$$

acts on an arbitrary predicate and arbitrary endo-function on nat

$$\begin{aligned} \text{eval} : (A \Rightarrow B) \times A \rightarrow B \\ (f, x) \mapsto f(x) \end{aligned}$$

$$\begin{aligned} \lambda p . (\pi_1(p))(\pi_2(p)) : (A \rightarrow B) \times A \rightarrow B \\ \pi_1(p) : (A \rightarrow C) \\ \pi_2(p) : A \end{aligned}$$

$$\begin{aligned} \text{comp} : (B \Rightarrow C) \times (A \Rightarrow B) \rightarrow (A \Rightarrow C) \\ (g, f) \mapsto g \circ f \end{aligned}$$

$$\begin{aligned} \lambda f . \lambda x . (\pi_1(f))(\pi_2(f)(x)) : ((B \rightarrow C) \times (A \rightarrow B)) \rightarrow (A \rightarrow C) \\ \pi_1(f) : (B \rightarrow C) \\ \pi_2(f) : (A \rightarrow B) \\ \pi_2(f)(x) : B \\ (\pi_1(f))(\pi_2(f)(x)) : C \\ \lambda x . (\pi_1(f))(\pi_2(f)(x)) : A \rightarrow C \end{aligned}$$

Note the observable behaviour is about when values get returned

this is what we care about!

# What does programming language theory study?

We want programs that are:  
**efficient, fast, and correct**

We ask:

- (1) How should we think about programs?
- (2) When are programs interchangeable?

terms in some version of simply-typed  $\lambda$ -calculus

Two notions of equality:

- (1) “equality as **functions**”
- (2) “equality as **programs**”

= same behaviour no matter  
what program you put them into

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$$P =_{\beta\eta} Q \implies P \simeq_{\text{ctx}} Q$$

converse is false!

**$\beta\eta$ -equality**  $=_{\beta\eta}$ : the congruence generated by  $\rightsquigarrow_\beta \cup \rightsquigarrow_\eta$

**observational equivalence**:  $P \simeq_{\text{obs}} Q$  iff whatever program  $C[\_]$  of type bool or nat we put them in,  $C[P]$  and  $C[Q]$  have the same behaviour

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$$(\lambda x . P)(Q) =_{\beta\eta} P[x \mapsto Q]$$

$$P =_{\beta\eta} \lambda x . P(x)$$

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$$P =_{\beta\eta} \langle \pi_1(P), \pi_2(P) \rangle$$

for every program  $C[\_]$  with a ‘hole’ such that  $C[P], C[Q] : \text{nat}$  or  $C[P], C[Q] : \text{bool}$ , we have  
 $C[P]$  terminates with output  $V$  and effect  $E \iff C[Q]$  terminates with output  $V$  and effect  $E$

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use the syntax and the  $\rightsquigarrow$  relations directly;  
generally inductive arguments

easy to refute observational equivalences;  
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Two schools:

- (1) **Syntactic techniques**
- (2) **Semantic techniques**

build semantic models and study those instead

use the syntax and the  $\rightsquigarrow$  relations directly; generally inductive arguments

easy to refute observational equivalences; hard to prove them!

easier to prove observational equivalences; hard to refute them!

# Coming up next

1. Introduce an idealised functional programming language
  2. Explain its semantic interpretation in CCCs
- 
3. Introduce differentiable programming
  4. Explain the interpretation in Diff

# Cartesian closed categories (cccs)

def: a cartesian closed category  $(\mathbb{C}, \times, 1, \Rightarrow)$  is a category  $\mathbb{C}$   
with finite products  $(\times, 1)$   
and a right adjoint  $A \Rightarrow (-)$  for every  $(-) \times A$

$$\mathbb{C}(X, A_1 \times A_2) \cong \mathbb{C}(X, A_1) \times \mathbb{C}(X, A_2)$$

$$f \mapsto (\pi_1 \circ f, \pi_2 \circ f)$$

$$\langle f_1, f_2 \rangle \leftarrow (f_1, f_2)$$

$$\langle f_1, f_2 \rangle(x) = (f_1 x, f_2 x)$$

$$\mathbb{C}(X \times A, B) \cong \mathbb{C}(X, A \Rightarrow B)$$

$$f \mapsto \Lambda(f) \quad \Lambda(f)(x) = f(x, \_)$$

$$\text{eval} \circ (f \times A) \leftarrow f$$

$$\tilde{f}(x, a) = f(x)(a)$$

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$$f \mapsto (\pi_1 \circ f, \pi_2 \circ f) \quad \text{projections}$$

pairing  $\langle f_1, f_2 \rangle \leftrightarrow (f_1, f_2)$

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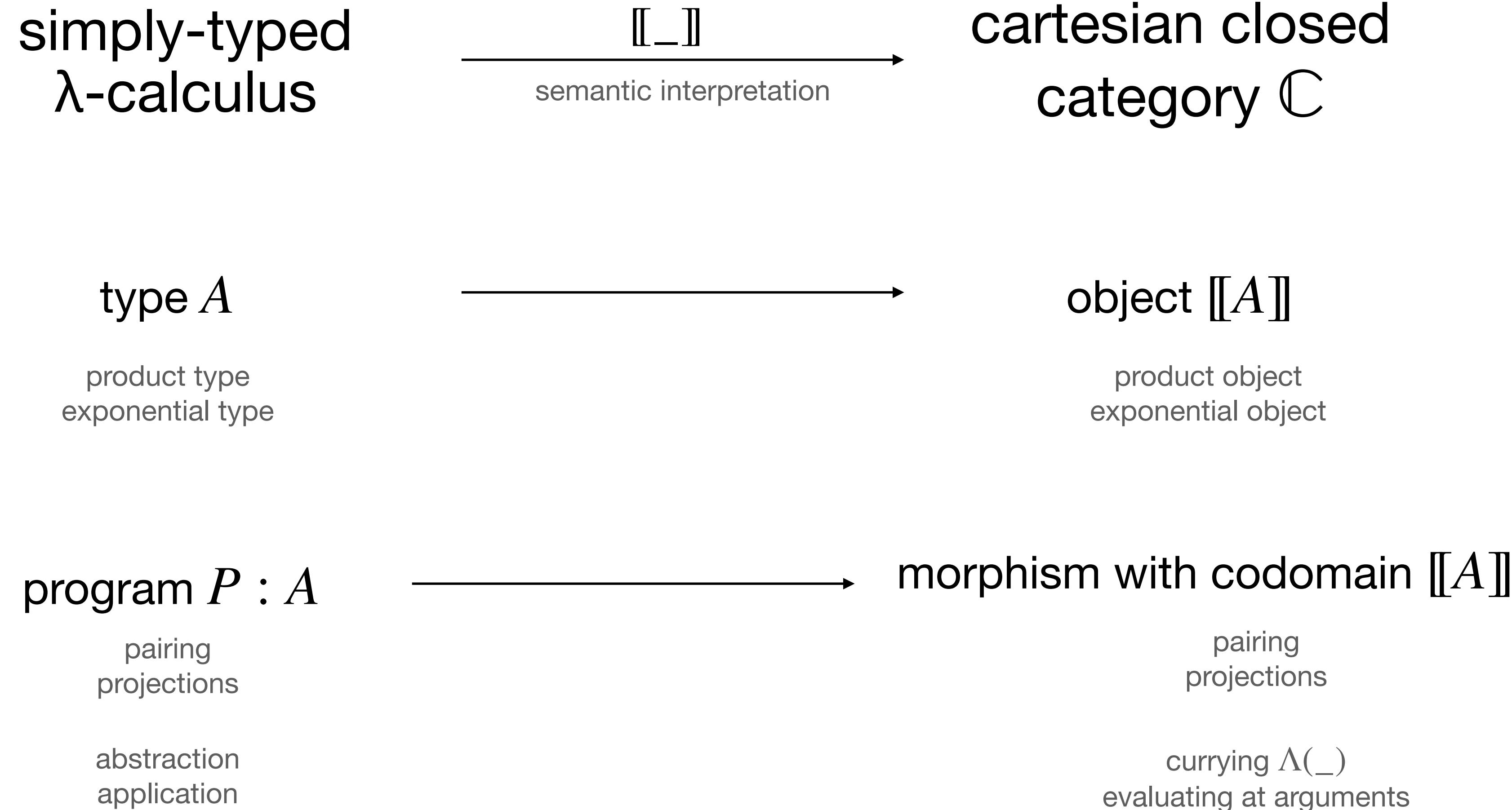
$$f \mapsto \Lambda(f) \quad \Lambda(f) = \lambda x . f(x, \_) \quad \text{abstraction}$$

$$\text{eval} \circ (f \times A) \leftrightarrow f$$

$$\text{eval} = \lambda(f, x) . f(x) \quad \text{application}$$

$$\text{eval} \circ (f \times A) = \lambda(x, a) . f(x)(a)$$

# Semantic interpretation



# Meanings for types in a CCC

Types  $\exists A, B ::= \text{nat} \mid \text{bool} \mid A \times B \mid A \rightarrow B$

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chosen objects  
eg  $2 := 1 + 1$ ,  $\mathbb{N} :=$  a natural numbers object

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$$[\![A \rightarrow B]\!] := ([\![A]\!] \Rightarrow [\![B]\!])$$

chosen objects

eg  $2 := 1 + 1$ ,  $\mathbb{N} :=$  a natural numbers object

$$[\![\text{nat} \rightarrow \text{bool}]\!] := (\mathbb{N} \Rightarrow 2)$$

$$[\![\text{bool} \rightarrow \text{bool}]\!] := (2 \Rightarrow 2)$$

⋮

# Meanings for terms in a CCC

## handling free variables

no free variables

**plus : nat × nat → nat**

assigns something of type nat whenever we give  $P : \text{nat} \times \text{nat}$

so  $\llbracket \text{plus} \rrbracket$  is a map  $\llbracket \text{nat} \rrbracket \times \llbracket \text{nat} \rrbracket \rightarrow \llbracket \text{nat} \rrbracket$ ;  
equivalently, a map  $1 \rightarrow ((\llbracket \text{nat} \rrbracket \times \llbracket \text{nat} \rrbracket) \Rightarrow \llbracket \text{nat} \rrbracket)$

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$b, n$  and  $m$  free

$\text{if}(b, n, m) : \text{nat}$

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$P : B$  with free variables  $(x_i : A_i)_{i=1,\dots,n}$  has  
interpretation  $\llbracket P \rrbracket : \prod_{i=1}^n \llbracket A_i \rrbracket \rightarrow \llbracket B \rrbracket$

assigns  $a_i \in \llbracket A_i \rrbracket$  to each  $x_i : A_i$   
eg  $\llbracket \text{if} \rrbracket(0,2,3) = 2$

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for  $P : B$  with no free variables,  
 $\llbracket P \rrbracket : 1 \rightarrow \llbracket B \rrbracket$

so eg  $P : A \rightarrow B$  is identified with an element of  $(\llbracket A \rrbracket \Rightarrow \llbracket B \rrbracket)$

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Fix an interpretation  $\llbracket c \rrbracket$  for each primitive  $c$

For  $P : B$  with no free variables,  $\llbracket P \rrbracket \in \llbracket B \rrbracket$ :

$$\llbracket \pi_i(P) \rrbracket = (\text{ith projection out } \llbracket P \rrbracket) \in \llbracket B_i \rrbracket$$

$$\llbracket \langle P_1, P_2 \rangle \rrbracket = (\llbracket P_1 \rrbracket, \llbracket P_2 \rrbracket) \in \llbracket B_1 \rrbracket \times \llbracket B_2 \rrbracket$$

$$\llbracket P(Q) \rrbracket = (\llbracket P \rrbracket) (\llbracket Q \rrbracket) \in \llbracket C \rrbracket$$

$$\llbracket \lambda x . P \rrbracket = \lambda b . \llbracket P \rrbracket(b) \in \llbracket B \rrbracket \Rightarrow \llbracket C \rrbracket$$

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$$\llbracket P(Q) \rrbracket(\vec{a}) = (\llbracket P \rrbracket(\vec{a})) (\llbracket Q \rrbracket(\vec{a})) \in \llbracket C \rrbracket$$

$$\llbracket \lambda x . P \rrbracket(\vec{a}) = \lambda b . \llbracket P \rrbracket(\vec{a}, b) \in \llbracket B \rrbracket \Rightarrow \llbracket C \rrbracket$$

# Meanings for terms in a CCC

$P : B$  with free variables  $(x_i : A_i)_{i=1,\dots,n}$   
has interpretation  
 $\llbracket P \rrbracket : \prod_{i=1}^n \llbracket A_i \rrbracket \rightarrow \llbracket B \rrbracket$

assigns  $a_i \in \llbracket A_i \rrbracket$  to each  $x_i : A_i$

Fix an interpretation  $\llbracket c \rrbracket$  for each primitive  $c$

Then:

$$\llbracket \pi_i(P) : B_i \rrbracket = \pi_i \circ \llbracket P : B_1 \times B_2 \rrbracket$$

$$\llbracket \langle P_1, P_2 \rangle : B_1 \times B_2 \rrbracket = \langle \llbracket P_1 : B_1 \rrbracket, \llbracket P_2 : B_2 \rrbracket \rangle \quad \langle f_1, f_2 \rangle(x) = (f_1 x, f_2 x)$$

$$\llbracket P(Q) : C \rrbracket = \text{eval} \circ \langle \llbracket P : B \rightarrow C \rrbracket, \llbracket Q : B \rrbracket \rangle$$

$$\llbracket \lambda x . P : B \rightarrow C \rrbracket = \Lambda(\llbracket P : C \rrbracket) \quad \Lambda(f) = \lambda x . f(x, \_)$$

$$\text{eval} \circ (f \times A) = \lambda(x, a) . f(x)(a)$$

# Soundness of the interpretation

$P : B$  with free variables  $(x_i : A_i)_{i=1,\dots,n}$   
has interpretation  
 $\llbracket P \rrbracket : \prod_{i=1}^n \llbracket A_i \rrbracket \rightarrow \llbracket B \rrbracket$

assigns  $a_i \in \llbracket A_i \rrbracket$  to each  $x_i : A_i$

for any CCC  $\mathbb{C}$  and any choice of base types and constants,

$$P =_{\beta\eta} Q \implies \llbracket P \rrbracket = \llbracket Q \rrbracket$$

in an **adequate** model,  $\llbracket P \rrbracket = \llbracket Q \rrbracket \implies P \simeq_{\text{obs}} Q$

in fact, simply-typed  $\lambda$ -calculus modulo  $=_{\beta\eta}$  is  
a sound and complete logic for CCCs

# What does programming language theory study?

We want programs that are:  
**efficient, fast, and correct**

We ask:

- (1) How should we think about programs?
- (2) When are programs interchangeable?

Two notions of equality:

$\beta\eta$ -equality  $=_{\beta\eta}$ : the congruence generated by  $\rightsquigarrow_\beta \cup \rightsquigarrow_\eta$

observational equivalence:  $P \simeq_{\text{obs}} Q$  iff whatever program  $C[_]$  of type bool or nat we put them in,  $C[P]$  and  $C[Q]$  have the same behaviour

# What does denotational semantics study?

We want programs that are:  
**efficient, fast, and correct**

We ask:

- (1) How should we think about programs?
- (2) When are programs interchangeable?

terms in some version of  
simply-typed  $\lambda$ -calculus  
interpreted in a CCC

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observational equivalence:  $P \simeq_{\text{obs}} Q$  iff whatever program  $C[_]$  of type bool or nat we put them in,  $C[P]$  and  $C[Q]$  have the same behaviour

use **adequate** models to reason about observational equivalence of programs

adequacy:  $\llbracket P \rrbracket = \llbracket Q \rrbracket \implies P \simeq_{\text{obs}} Q$

# Some example interpretations

languages with no effects

plain CCCs

languages with printing,  
global memory, exceptions

CCCs with a (strong)  
monad

the monad  $T$  describes the  
effect, eg  $(-)+1$  or  $S^* \times (-)$

languages with local memory

presheaf  
categories

think: programs parametrised by  
possible states of the memory

languages with recursion

order-enriched  
categories

$\leq$  models ‘how defined’ a function is

each recursive call goes up the order;  
the whole loop is then a fixpoint

looping forever modelled by a bottom element

# How should we think about programs?

```
fun add(x, y):  
    return (x + y)
```

a function  $\mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$

```
fun divide(x, y):  
    return (x / y)
```

a function  $\mathbb{Z} \times \mathbb{Z}_{\neq 0} \rightarrow \mathbb{Q}$

a function  $\mathbb{Z} \times \mathbb{Z} \rightarrow \mathbb{Q} + \{\text{fail}\}$

```
fun print_and_return(x):  
    print "hello";  
    return x;
```

a function  $\mathbb{N} \rightarrow \{a, b, \dots, z\}^* \times \mathbb{N}$   
 $x \mapsto (\text{hello}, x)$

```
let b = flip(p);  
return b;
```

a probability distribution on  
 $\{\text{true}, \text{false}\}$

```
normalise(  
    let x = sample(bernoulli(0.8));  
    let r = (if x then 10 else 3);  
    observe 0.45 from exponential(r)  
    return(x)  
)
```

some measurable function (??)

# Coming up next

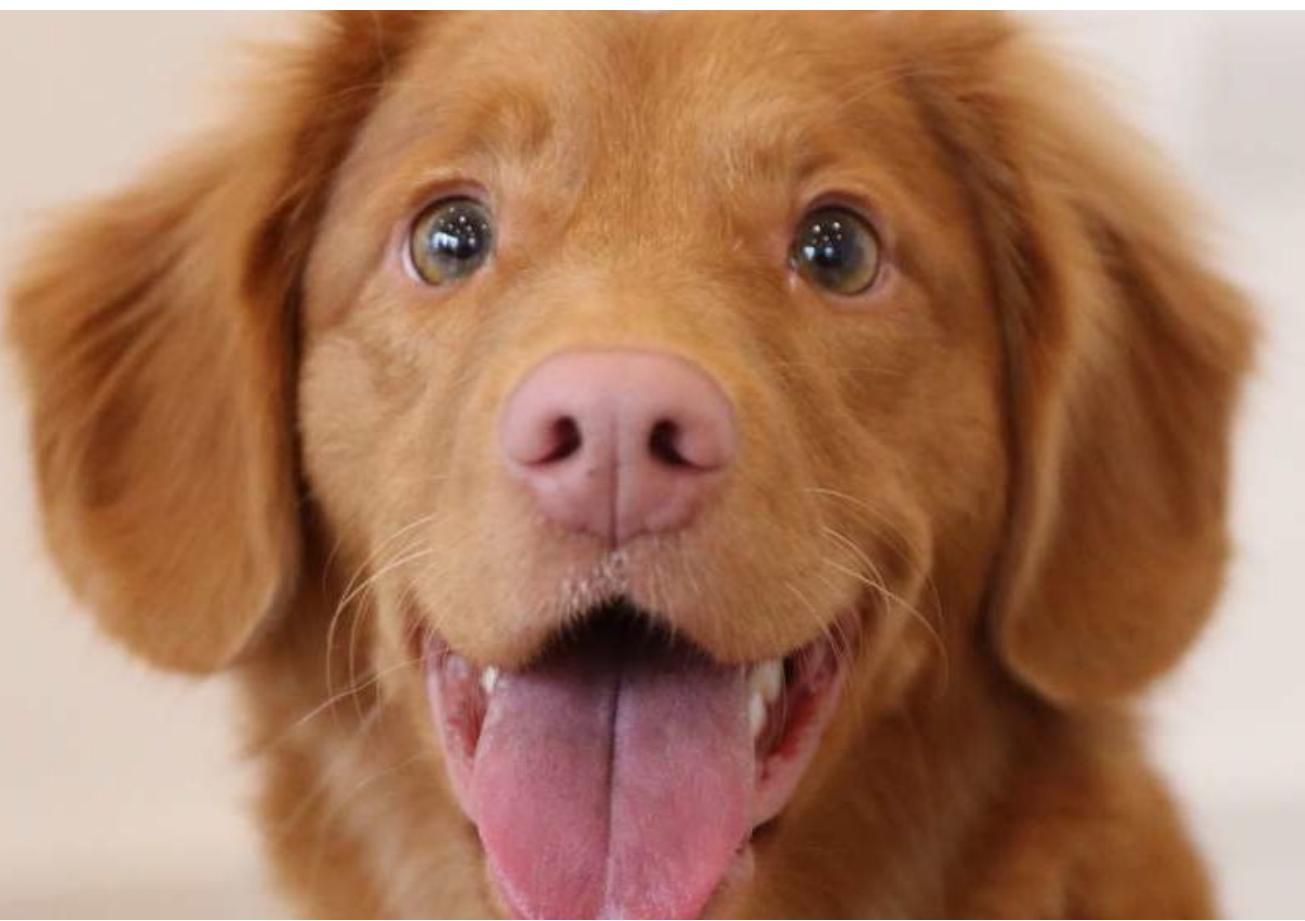
1. Introduce an idealised functional programming language
2. Explain its semantic interpretation in CCCs

- 
3. Introduce differentiable programming
  4. Explain the interpretation in Diff

high-dimensional  
input



→ cat



→ dog

high-dimensional  
input

program  $P$  with many parameters

---

e.g. a neural network with many layers, and different weights for the activation functions

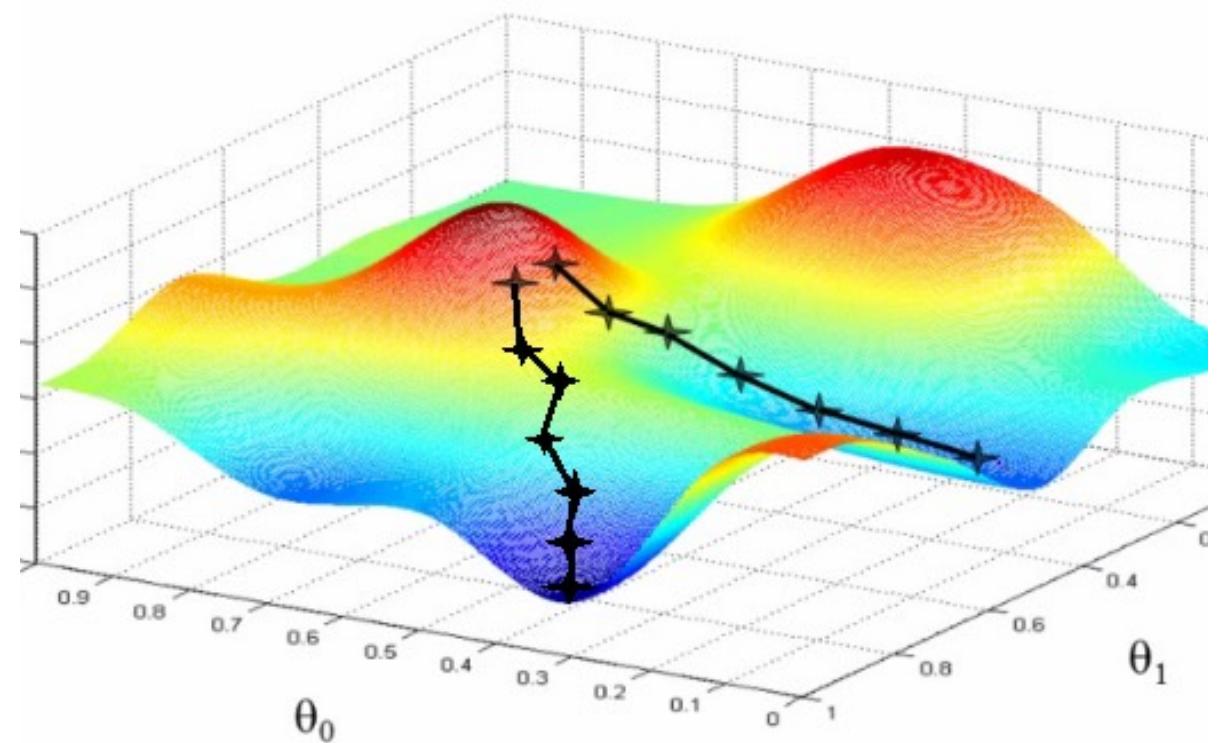
low-dimensional  
output

high-dimensional  
input

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low-dimensional  
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ie **differentiate** the function described by  $P$

aim: optimise the parameters for  $P$

so that, eg, it classifies cats as cats as often as possible

can be done  
numerically, but it's  
hard in general!

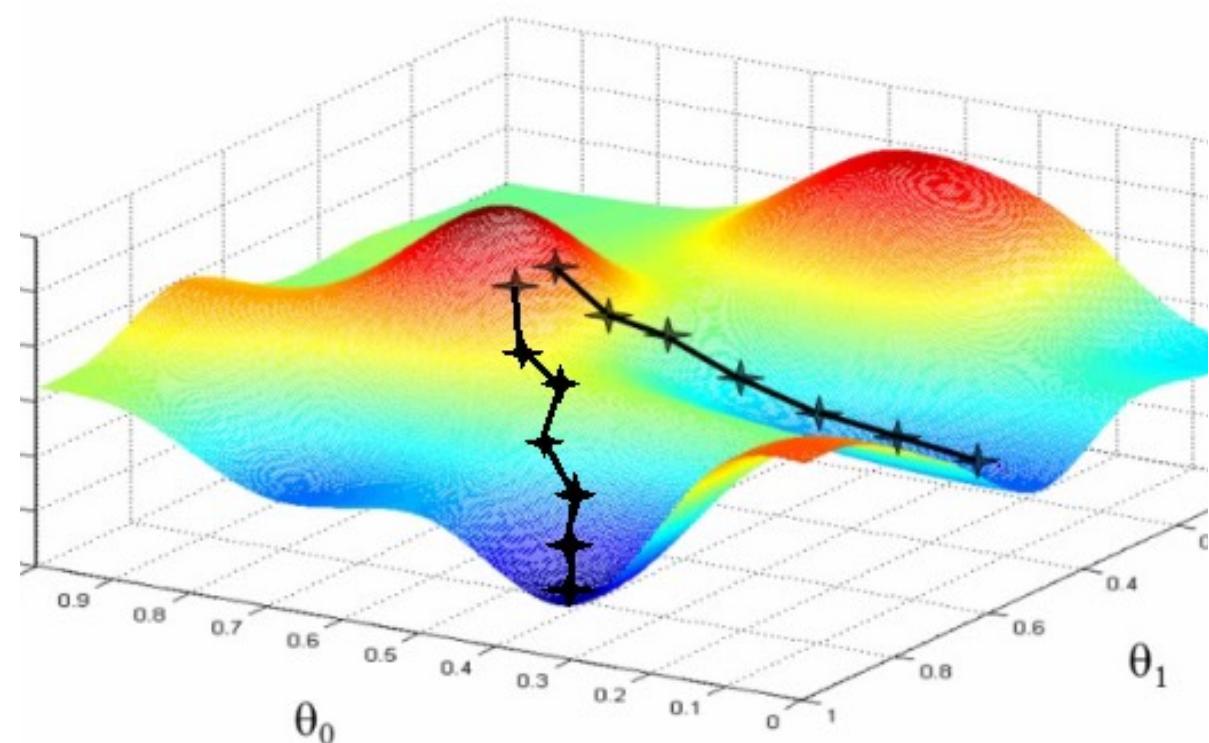
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can we write an algorithm to calculate derivatives exactly?

...and can we prove this is correct?

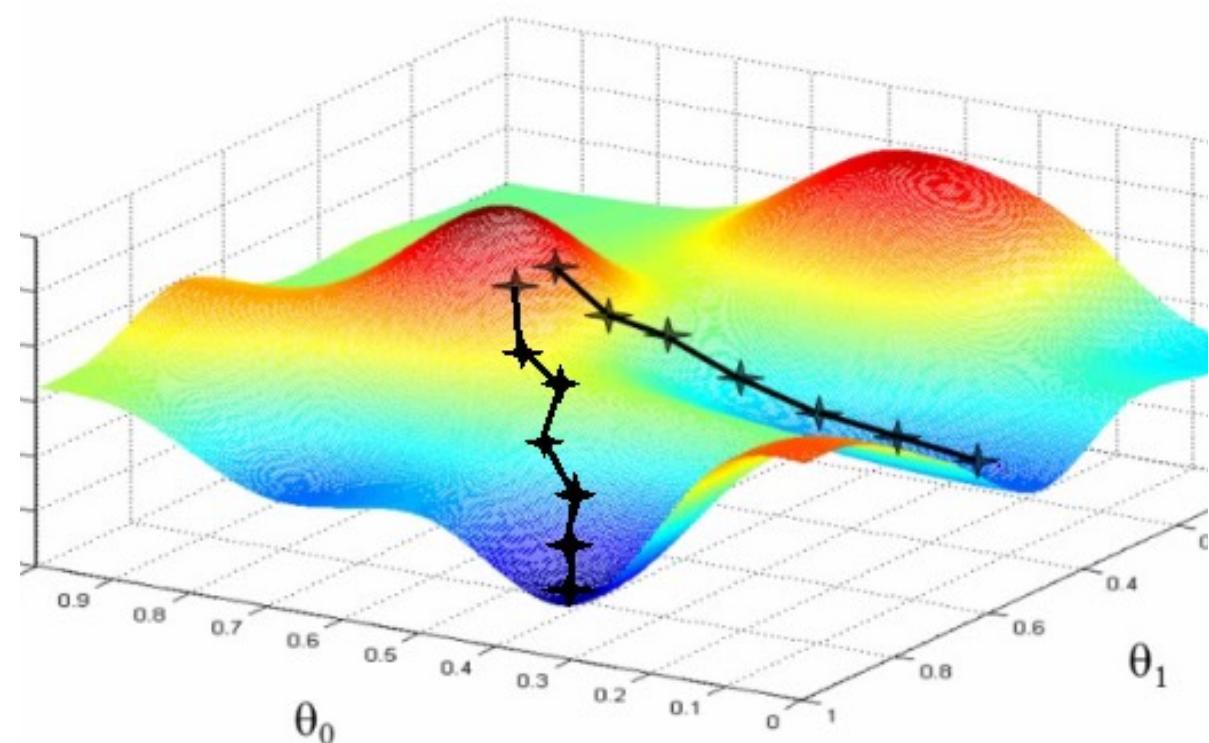
“forward AD”,  
“reverse AD”, etc

high-dimensional  
input

program  $P$  with many parameters

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output



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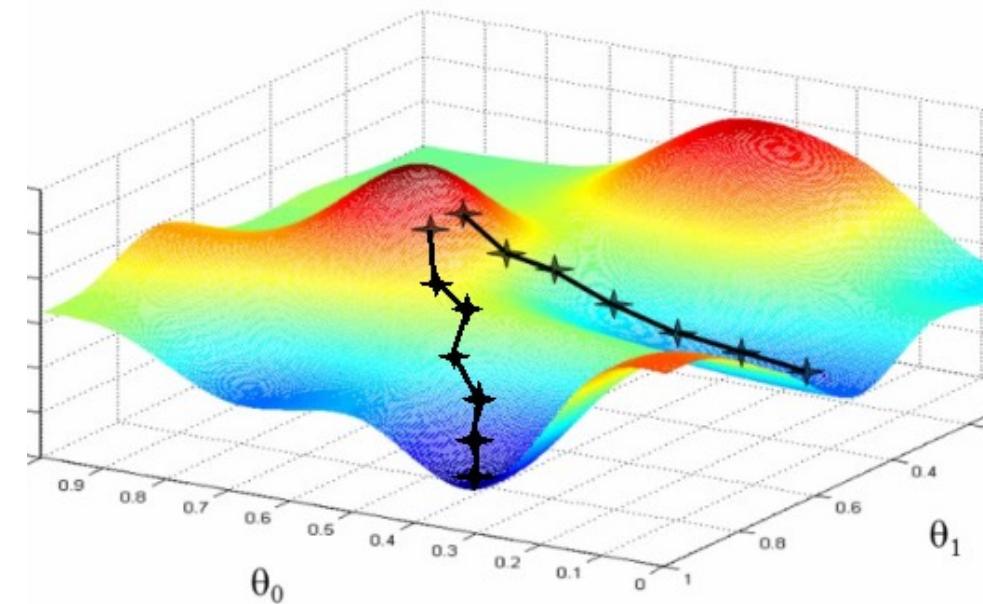
...and can we prove this is **correct**?

“forward AD”,  
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**differentiable programming** (TensorFlow, PyTorch, etc)

– languages where you can automatically  
compute the derivative of any program

high-dimensional  
input



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= a serious bottleneck

from the denotational semantics POV:

(1)  $\llbracket P \rrbracket$  is some smooth function  $\mathbb{R}^n \rightarrow \mathbb{R}$

(2) aim: to algorithmically define a program  $D(P)$

and prove that  $\llbracket D(P) \rrbracket = D(\llbracket P \rrbracket)$

essentially, using the chain rule  
and “dual numbers”

# Proving correctness of automatic differentiation

[Huot, Staton, Vakar]

a natural suggestion:

- (1) we only care about the programs returning a value, ie those of type `real`
- (2) take simply-typed  $\lambda$ -calculus + primitives for real numbers etc
- (3) a program  $P : \text{real}$  is meant to represent a smooth function  $\mathbb{R}^n \rightarrow \mathbb{R}$
- (4) define  $D(P)$  by induction on the simply-typed  $\lambda$ -calculus  
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$$\begin{aligned}\vec{\mathcal{D}}(x) &\stackrel{\text{def}}{=} x & \vec{\mathcal{D}}(\underline{c}) &\stackrel{\text{def}}{=} \langle \underline{c}, 0 \rangle \\ \vec{\mathcal{D}}(t + s) &\stackrel{\text{def}}{=} \mathbf{case} \vec{\mathcal{D}}(t) \mathbf{of} \langle x, x' \rangle \rightarrow \mathbf{case} \vec{\mathcal{D}}(s) \mathbf{of} \langle y, y' \rangle \rightarrow \langle x + y, x' + y' \rangle \\ \vec{\mathcal{D}}(t * s) &\stackrel{\text{def}}{=} \mathbf{case} \vec{\mathcal{D}}(t) \mathbf{of} \langle x, x' \rangle \rightarrow \mathbf{case} \vec{\mathcal{D}}(s) \mathbf{of} \langle y, y' \rangle \rightarrow \langle x * y, x * y' + x' * y \rangle \\ \vec{\mathcal{D}}(\varsigma(t)) &\stackrel{\text{def}}{=} \mathbf{case} \vec{\mathcal{D}}(t) \mathbf{of} \langle x, x' \rangle \rightarrow \mathbf{let} y = \varsigma(x) \mathbf{in} \langle y, x' * y * (1 - y) \rangle \\ \vec{\mathcal{D}}(\lambda x. t) &\stackrel{\text{def}}{=} \lambda x. \vec{\mathcal{D}}(t) & \vec{\mathcal{D}}(t s) &\stackrel{\text{def}}{=} \vec{\mathcal{D}}(t) \vec{\mathcal{D}}(s) & \vec{\mathcal{D}}(\langle t_1, \dots, t_n \rangle) &\stackrel{\text{def}}{=} \langle \vec{\mathcal{D}}(t_1), \dots, \vec{\mathcal{D}}(t_n) \rangle\end{aligned}$$

from *Correctness of Automatic Differentiation via Diffeologies and Categorical Gluing*

# Proving correctness of automatic differentiation

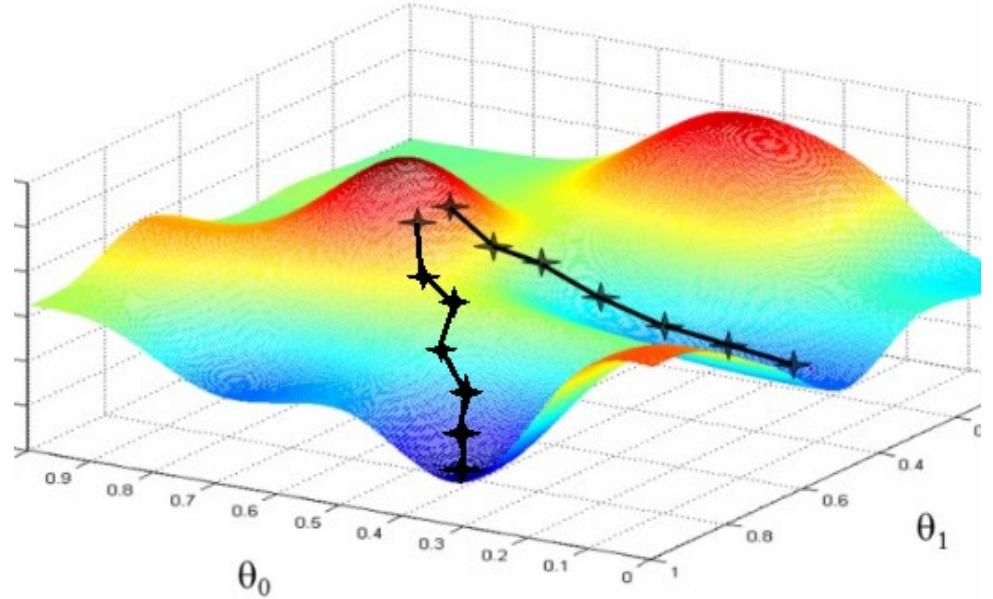
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ie. we interpret in the category of cartesian spaces ( $= \mathbb{R}^n$  for some  $n$ ) and smooth maps  
**but this category is not cartesian closed!** and even  $P : \text{real}$  may contain lambdas,  
eg  $(\lambda f. \lambda x. f(x + x))(\exp)(2)$

high-dimensional  
input



[https://www.youtube.com/  
watch?v=5u4G23\\_OohI](https://www.youtube.com/watch?v=5u4G23_OohI)

## program $P$ with many parameters

eg a neural network with many layers, and different weights for the activation functions

low-dimensional  
output

ie **differentiate** the function described by  $P$

**aim: optimise the parameters for  $P$**

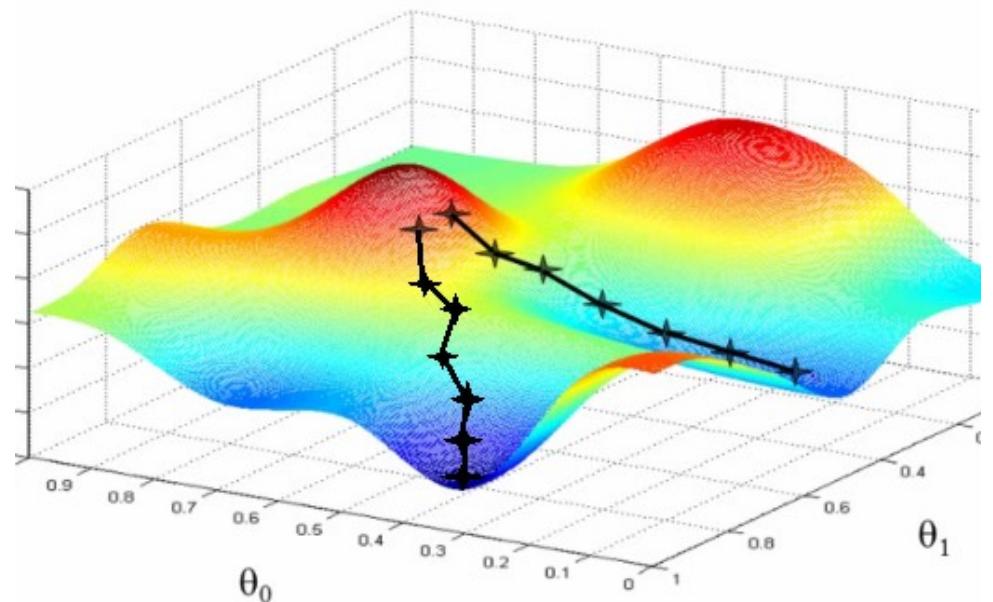
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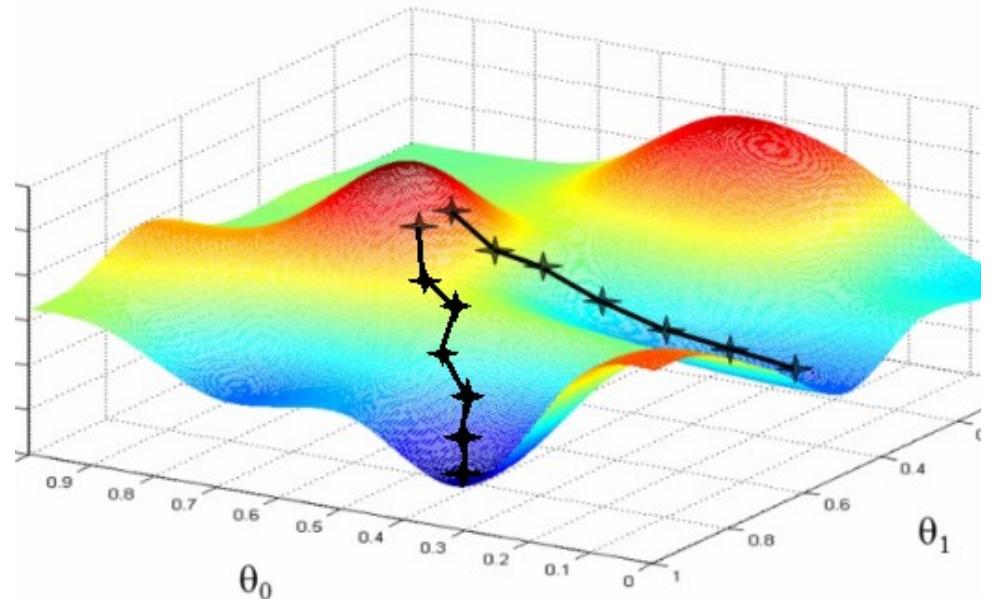
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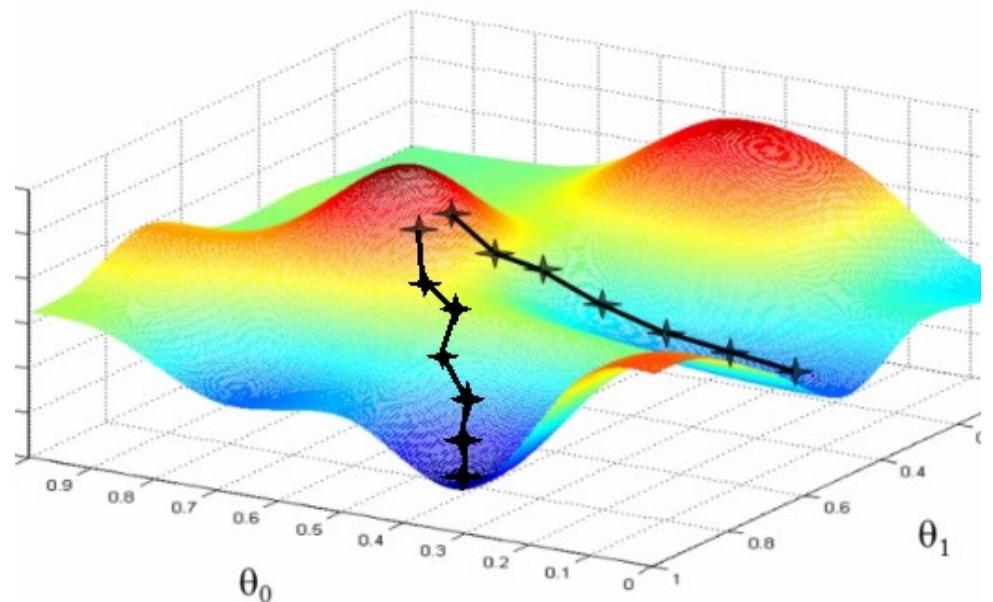
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we need a CCC that supports some notion of derivative



# The category of diffeological spaces

Diff is a nice semantic model! It has:

- (1) **cartesian closure** = can model product and function types
- (2) a full embedding  $\text{CartSp} \rightarrow \text{Diff}$  = conservativity over the natural model,  
good ways to interpret reals etc
- (3) **coproducts** = can interpret sum types (~ disjoint unions)
- (4) **initial algebras for endofunctors**  
= can interpret lists and similar inductive types

# Proving correctness of automatic differentiation

[Huot, Staton, Vakar]

The strategy:

- (a) interpret programs  $P$  in  $\text{Diff}$
- (b) prove that  $\llbracket P : \text{real} \rrbracket$  always lands in  $\text{CartSp}$ , even if it has lambdas
- (c) prove a correctness property for differentiation, at every type
- (d) deduce correctness of the  $D(-)$  algorithm at type  $\text{real}$

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And, at type real the interpretation coincides with the natural one

# Diff at work for semantics

- (1) An analogy
- (2) Adding recursion
- (3) Cutting down the model: full abstraction

# Probabilistic programming

Idea:

- (1) programs express statistical models,  
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How do we interpret probabilistic programs?  
What is a good semantic model?

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probabilistic programs  
'should' be interpreted by  
**measurable functions**

but Meas is not cartesian closed!  
= no way to interpret higher-order functions

How do we interpret probabilistic programs?  
What is a good semantic model?

# Quasi-Borel spaces

[Heunen, Kammar, Moss, Scibior, Staton, Vakar, Yang]

Diff = category of concrete sheaves on cartesian manifolds

QBS = category of concrete sheaves on standard Borel spaces

always a quasi-topos, in particular a CCC

QBS provides a good semantic model for probabilistic programming,  
just as Diff provides a good semantic model for differentiable programming

# Diff at work for semantics

- (1) An analogy: quasi-Borel spaces [Heunen, Kammar, Moss, Scibior, Staton, Vakar, Yang]
- (2) Adding recursion
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# Adding recursion to simply-typed $\lambda$ -calculus

[Scott, Plotkin,...]

$\text{plus} : \mathbf{N} \times \mathbf{N} \rightarrow \mathbf{N}$  is the least map satisfying

$$\text{plus}(x, 0) = x$$

$$\text{plus}(x, y + 1) = \text{plus}(x, y) + 1$$

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recursion in simply-typed  $\lambda$ -calculus:

$$\frac{P : A \rightarrow A}{\text{fix}(M) : A}$$

$$\text{fix}(M) \rightsquigarrow M(\text{fix}(M))$$

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# Adding recursion to simply-typed $\lambda$ -calculus

[Scott, Plotkin,...]

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by Tarski's fixpoint theorem

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Can extend correctness results for AD to languages with recursion!

# Diff at work for semantics

- (1) An analogy: quasi-Borel spaces [Heunen, Kammar, Moss, Scibior, Staton, Vakar, Yang]
- (2) Adding recursion [Vakar] [Vakar, Kammar, Staton]
- (3) Cutting down the model: full abstraction

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[Kammar, Katsumata, S.]

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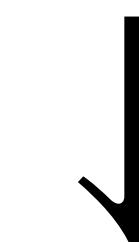
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objects = diffeological spaces paired with a family of relations  
morphisms = smooth maps preserving the relations

(new model)



preserves primitives and products,  
but **not** exponentials

choose the class of relations intensionally  
so maps preserving the relation are definable

Diff

# Diff at work for semantics

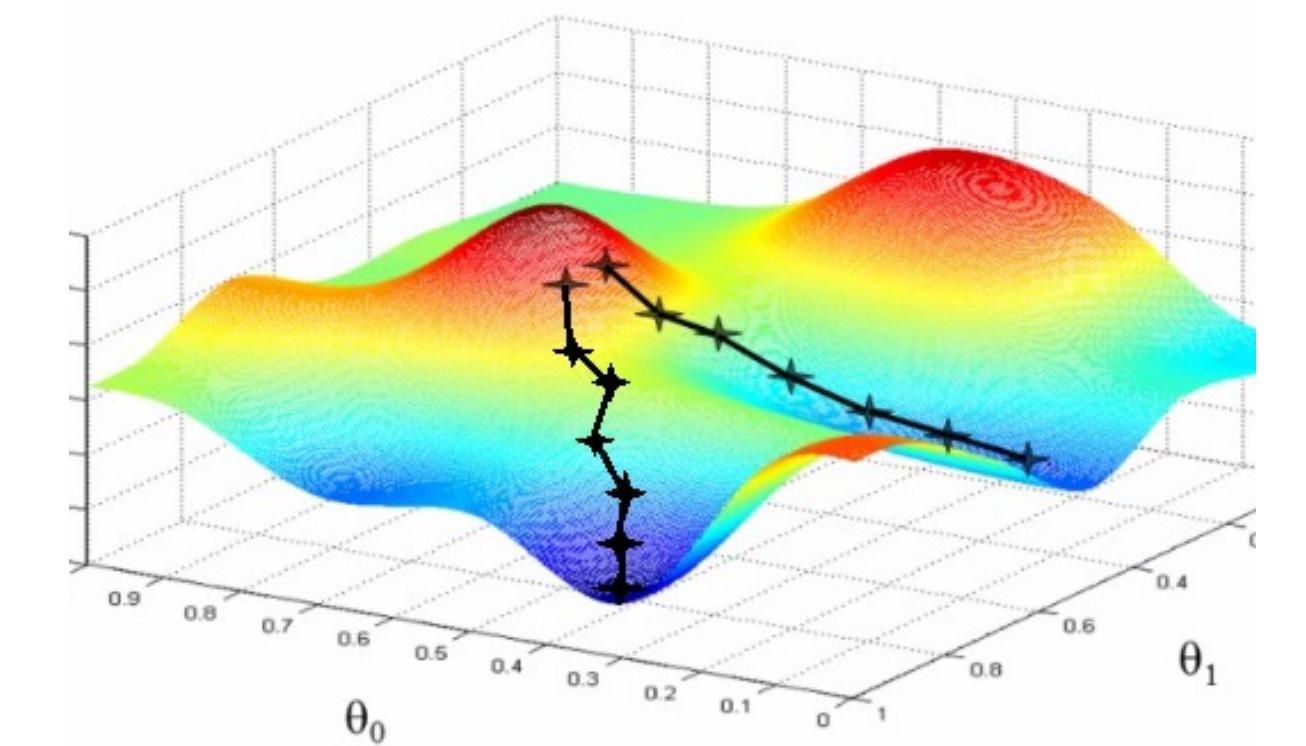
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## Denotational semantics:

- idealised functional programming language
  - = simply-typed  $\lambda$ -calculus (+ extensions)
- interchangeability of programs
  - = observational equivalence finer than equality-on-arguments!
- interpret programs in CCCs (+ extensions)

$$P \mapsto D(P)$$

Diff is a good model for studying  
automatic differentiation of programs



[https://www.youtube.com/watch?  
v=5u4G23\\_Oohl](https://www.youtube.com/watch?v=5u4G23_Oohl)