

Simple essence of AD

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24 de Maio

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Definition of Derivative

Definition

A function $f : \mathbb{R} \rightarrow \mathbb{R}$ is differentiable at $x \in \mathbb{R}$, if there exists a number $f'(x)$ such that:

$$\lim_{\varepsilon \rightarrow 0} \frac{f(x + \varepsilon) - f(x)}{\varepsilon} = f'(x)$$

$$\lim_{\varepsilon \rightarrow 0} \frac{f(x + \varepsilon) - f(x)}{\varepsilon} - f'(x) = 0 \Leftrightarrow \lim_{\varepsilon \rightarrow 0} \frac{f(x + \varepsilon) - f(x) - (\varepsilon \cdot f'(x))}{\varepsilon} = 0$$

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Definition

A function $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$ is differentiable at $x \in \mathbb{R}^m$, if there exists a unique linear transformation $\mu : \mathbb{R}^m \rightarrow \mathbb{R}^n$ such that:

$$\lim_{\varepsilon \rightarrow 0} \frac{\|f(x + \varepsilon) - f(x) - \mu(\varepsilon)\|}{\|\varepsilon\|} = 0$$

Derivate as a linear map

Definition

Let $f :: a \rightarrow b$ be a function, where a and b are vectorial spaces that share a common underlying field. The first derivative definition is the following:

$$\mathcal{D} :: (a \rightarrow b) \rightarrow (a \rightarrow (a \multimap b))$$

If we differentiate two times, we have:

$$\mathcal{D}^2 = \mathcal{D} \circ \mathcal{D} :: (a \rightarrow b) \rightarrow (a \rightarrow (a \multimap a \multimap b))$$

Rules for Differentiation - Sequential Composition

Theorem

Let $f :: a \rightarrow b$ and $g :: b \rightarrow c$ be two functions. Then the derivative of the composition of f and g is:

$$\mathcal{D} (g \circ f) a = \mathcal{D} g (f a) \circ \mathcal{D} f a$$

Rules for Differentiation - Sequential Composition

Unfortunately the previous theorem isn't a efficient recipe for composition. As such we will introduce a second derivative definition:

$$\begin{aligned}\mathcal{D}_0^+ &:: (a \rightarrow b) \rightarrow ((a \rightarrow b) \times (a \rightarrow (a \multimap b))) \\ \mathcal{D}_0^+ f &= (f, \mathcal{D} f)\end{aligned}$$

With this, the chain rule will have the following expression:

$$\begin{aligned}\mathcal{D}_0^+ (g \circ f) & \\ \{\text{definition of } \mathcal{D}_0^+\} & \\ = (g \circ f, \mathcal{D} (g \circ f)) & \\ \{\text{theorem and definition of } g \circ f\} & \\ = (\lambda a \rightarrow g(f a), \lambda a \rightarrow \mathcal{D} g (f a) \circ \mathcal{D} f a) &\end{aligned}$$

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Rules for Differentiation - Sequential Composition

Having in mind optimizations, we introduce the third and last derivative definition:

$$\begin{aligned}\mathcal{D}^+ &:: (a \rightarrow b) \rightarrow (a \rightarrow (b \times (a \multimap b))) \\ \mathcal{D}^+ f a &= (f a, \mathcal{D} f a)\end{aligned}$$

As \times has more priority than \rightarrow and \multimap , we can rewrite \mathcal{D}^+ as:

$$\begin{aligned}\mathcal{D}^+ &:: (a \rightarrow b) \rightarrow (a \rightarrow b \times (a \multimap b)) \\ \mathcal{D}^+ f a &= (f a, \mathcal{D} f a)\end{aligned}$$

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Rules for Differentiation - Sequential Composition

Corollary

\mathcal{D}^+ is (efficiently) compositional with respect to (\circ) , that is, in Haskell:

$$\mathcal{D}^+ (g \circ f) a = \mathbf{let} \{ (b, f') = \mathcal{D}^+ f a; (c, g') = \mathcal{D}^+ g b \} \\ \mathbf{in} (c, g' \circ f')$$

$$\begin{array}{ccccc} (C \times C^B) \times B^A & \xleftarrow{\mathcal{D}^+ g \times id} & B \times B^A & \xleftarrow{\mathcal{D}^+ f} & A \\ \downarrow (id \times (uncurry (\circ))) \circ assocr & & & \swarrow \mathcal{D}^+(g \circ f) & \\ C \times C^A & & & & \end{array}$$

Rules for Differentiation - Parallel Composition

Another important way of combining functions is the operation cross, that combines two functions in parallel:

$$\begin{aligned} (\times) &:: (a \rightarrow c) \rightarrow (b \rightarrow d) \rightarrow (a \times b \rightarrow c \times d) \\ f \times g &= \lambda(a, b) \rightarrow (f \ a, g \ b) \end{aligned}$$

Theorem

Let $f :: a \rightarrow c$ and $g :: b \rightarrow d$ be two function. Then the cross rule is the following:

$$\mathcal{D}(f \times g)(a, b) = \mathcal{D}f \ a \times \mathcal{D}g \ b$$

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$$\mathcal{D}(f \times g)(a, b) = \mathcal{D} f \ a \times \mathcal{D} g \ b$$

Rules for Differentiation - Parallel Composition

Corollary

The function \mathcal{D}^+ is compositional in relation to (\times)

$$\mathcal{D}^+ (f \times g) (a, b) = \mathbf{let} \{ (c, f') = \mathcal{D}^+ f a; (d, g') = \mathcal{D}^+ g b \} \\ \mathbf{in} ((c, d), f' \times g')$$

Derivative and Linear Functions

Definition

A function f is said to be linear when preserves addition and scalar multiplication.

$$f(a + a') = f a + f a'$$

$$f(s \cdot a) = s \cdot f a$$

Theorem

For all linear functions f , $\mathcal{D} f a = f$.

Corollary

For all linear functions f , $\mathcal{D}^+ f = \lambda a \rightarrow (fa, f)$.

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Corollary

For all linear functions f , $\mathcal{D}^+ f = \lambda a \rightarrow (fa, f)$.

A short introduction

- We want to calculate \mathcal{D}^+ .
- However, \mathcal{D} is not computable.
- Solution: reimplement corollaries using category theory.

Instance deduction

Using corollaries 3.1 and 1.1 we can determine that

- (DP.1) $\mathcal{D}^+ id = \lambda a \rightarrow (id\ a, id)$
- (DP.2)

$$\mathcal{D}^+ (g \circ f) = \lambda a \rightarrow \mathbf{let}\ \{(b, f') = \mathcal{D}^+ f\ a; (c, g') = \mathcal{D}^+ g\ b\} \\ \mathbf{in}\ (c, g' \circ f')$$

Saying that $\hat{\mathcal{D}}$ is a functor is equivalent to, for all f and g functions of appropriate types:

$$id = \hat{\mathcal{D}}\ id = \mathcal{D}\ (\mathcal{D}^+ id)$$

$$\hat{\mathcal{D}}\ g \circ \hat{\mathcal{D}}\ f = \hat{\mathcal{D}}\ (g \circ f) = \mathcal{D}\ (\mathcal{D}^+ (g \circ f))$$

Instance deduction

Based on (DP.1) and (DP.2) we'll rewrite the above into the following definition:

$$id = \mathcal{D} (\lambda a \rightarrow (id \ a, id))$$

$$\hat{\mathcal{D}} \ g \circ \hat{\mathcal{D}} \ f = \mathcal{D} (\lambda a \rightarrow \mathbf{let} \ \{(b, f') = \mathcal{D}^+ \ f \ a; (c, g') = \mathcal{D}^+ \ g \ b\} \ \mathbf{in} \ (c, g' \circ f'))$$

The first equation shown above has a trivial solution.

To solve the second we'll first solve a more general one:

$$\mathcal{D} \ g \circ \mathcal{D} \ f = \mathcal{D} (\lambda a \rightarrow \mathbf{let} \ \{(b, f') = f \ a; (c, g') = g \ b\} \ \mathbf{in} \ (c, g' \circ f'))$$

This condition also leads us to a trivial solution inside our instance.

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Instance deduction

$\hat{\mathcal{D}}$ definition for linear functions

$linearD :: (a \rightarrow b) \rightarrow \mathcal{D} a b$

$linearD f = \mathcal{D} (\lambda a \rightarrow (f a, f))$

Categorical instance we've deduced

instance *Category* \mathcal{D} **where**

$id = linearD id$

$\mathcal{D} g \circ \mathcal{D} f =$

$\mathcal{D} (\lambda a \rightarrow \mathbf{let} \{ (b, f') = f a; (c, g') = g b \} \mathbf{in} (c, g' \circ f'))$

Instance proof

In order to prove that the instance is correct we must check if it follows laws (C.1) and (C.2).

First we must make a concession: that we only use morfisms arising from \mathcal{D}^+ . If we do, then \mathcal{D}^+ is a functor.

(C.1) proof

$$\text{id} \circ \hat{\mathcal{D}} f$$

{ functor law for id (specification of $\hat{\mathcal{D}}$) }

$$= \hat{\mathcal{D}} \text{id} \circ \hat{\mathcal{D}} f$$

{ functor law for (\circ) }

$$= \hat{\mathcal{D}} (\text{id} \circ f)$$

{ categorical law }

$$= \hat{\mathcal{D}} f$$

Instance proof

(C.2) proof

$$\begin{aligned}
 & \hat{\mathcal{D}} h \circ (\hat{\mathcal{D}} g \circ \hat{\mathcal{D}} f) \\
 & \{ \text{2x functor law for } (\circ) \} \\
 & = \hat{\mathcal{D}} (h \circ (g \circ f)) \\
 & \{ \text{categorical law} \} \\
 & = \hat{\mathcal{D}} ((h \circ g) \circ f) \\
 & \{ \text{2x functor law for } (\circ) \} \\
 & = (\hat{\mathcal{D}} h \circ \hat{\mathcal{D}} g) \circ \hat{\mathcal{D}} f
 \end{aligned}$$

Note

Those proofs don't require anything from \mathcal{D} and $\hat{\mathcal{D}}$ aside from functor laws. As such, all other instances of categories created from a functor won't require further proving like this one did.

Monoidal categories and functors

Generalized parallel composition shall be defined using a monoidal category:

class *Category* $k \Rightarrow \text{Monoidal } k$ **where**

$(\times) :: (a \text{ ' } k' \text{ } c) \rightarrow (b \text{ ' } k' \text{ } d) \rightarrow ((a \times b) \text{ ' } k' \text{ } (c \times d))$

instance *Monoidal* (\rightarrow) **where**

$f \times g = \lambda(a, b) \rightarrow (f \text{ } a, g \text{ } b)$

Monoidal Functor definition

A monoidal functor F between categories \mathcal{U} and \mathcal{V} is such that:

- F is a functor
- $F(f \times g) = F f \times F g$

Instance deduction

From corollary 2.1 we can deduce that:

$\mathcal{D}^+ (f \times g) = \lambda(a, b) \rightarrow \mathbf{let} \{ (c, f') = \mathcal{D}^+ f a; (d, g') = \mathcal{D}^+ g b \}$
 $\mathbf{in} ((c, d), f' \times g')$

Deriving F from $\hat{\mathcal{D}}$ leaves us with the following definition:

$\mathcal{D} (\mathcal{D}^+ f) \times \mathcal{D} (\mathcal{D}^+ g) = \mathcal{D} (\mathcal{D}^+ (f \times g))$

Using the same method as before, we replace \mathcal{D}^+ with its definition and generalize the condition:

$\mathcal{D} f \times \mathcal{D} g =$

$\mathcal{D} (\lambda(a, b) \rightarrow \mathbf{let} \{ (c, f') = f a; (d, g') = g b \} \mathbf{in} ((c, d), f' \times g'))$

and this is enough for our new instance.

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Deriving F from $\hat{\mathcal{D}}$ leaves us with the following definition:

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and this is enough for our new instance.

Instance deduction

Categorical instance we've deduced

instance *Monoidal* \mathcal{D} **where**

$$\mathcal{D} f \times \mathcal{D} g = \mathcal{D} (\lambda(a, b) \rightarrow \mathbf{let} \{ (c, f') = f a; (d, g') = g b \} \\ \mathbf{in} ((c, d), f' \times g'))$$

Cartesian categories and functors

class *Monoidal* $k \Rightarrow \text{Cartesian } k$ **where**

$exl :: (a, b) \rightarrow k \rightarrow a$

$exr :: (a, b) \rightarrow k \rightarrow b$

$dup :: a \rightarrow k \rightarrow (a, a)$

instance *Cartesian* (\rightarrow) **where**

$exl = \lambda(a, b) \rightarrow a$

$exr = \lambda(a, b) \rightarrow b$

$dup = \lambda a \rightarrow (a, a)$

A cartesian functor F between categories \mathcal{U} and \mathcal{V} is such that:

- F is a monoidal functor
- $F \text{ exl} = \text{exl}$
- $F \text{ exp} = \text{exp}$
- $F \text{ dup} = \text{dup}$

Instance deduction

From corollary 3.1 and from exl , exr and dup being linear functions we can deduce that:

$$\mathcal{D}^+ exl = \lambda p \rightarrow (exl\ p, exl)$$

$$\mathcal{D}^+ exr = \lambda p \rightarrow (exr\ p, exr)$$

$$\mathcal{D}^+ dup = \lambda p \rightarrow (dup\ a, dup)$$

With this in mind we can arrive at our instance:

$$exl = \mathcal{D} (\mathcal{D}^+ exl)$$

$$exr = \mathcal{D} (\mathcal{D}^+ exr)$$

$$dup = \mathcal{D} (\mathcal{D}^+ dup)$$

Instance deduction

Replacing \mathcal{D}^+ with its definition and remembering linearD's definition we can obtain:

$exl = \text{linearD } exl$

$exr = \text{linearD } exr$

$dup = \text{linearD } dup$

and convert this directly into a new instance:

Categorical instance we've deduced

instance *Cartesian D* **where**

$exl = \text{linearD } exl$

$exr = \text{linearD } exr$

$dup = \text{linearD } dup$

Cocartesian category

This type of categories is the dual of the cartesian type of categories.

Note

In this article coproducts are categorical products, i.e., biproducts

Definition

```
class Category  $k \Rightarrow \text{Cocartesian } k$  where
  inl ::  $a \text{ ' } k \text{ ' } (a, b)$ 
  inr ::  $b \text{ ' } k \text{ ' } (a, b)$ 
  jam ::  $(a, a) \text{ ' } k \text{ ' } a$ 
```

Cocartesian functors

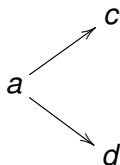
Cocartesian functor definition

A cocartesian functor F between categories \mathcal{U} and \mathcal{V} is such that:

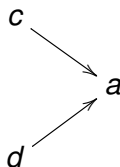
- F is a functor
- $F \text{ inl} = \text{inl}$
- $F \text{ inr} = \text{inr}$
- $F \text{ jam} = \text{jam}$

Fork and Join

- $\Delta :: \text{Cartesian } k \Rightarrow (a \text{ ' } k' \text{ } c) \rightarrow (a \text{ ' } k' \text{ } d) \rightarrow (a \text{ ' } k' \text{ } (c \times d))$



- $\nabla :: \text{Cartesian } k \Rightarrow (c \text{ ' } k' \text{ } a) \rightarrow (d \text{ ' } k' \text{ } a) \rightarrow ((c \times d) \text{ ' } k' \text{ } a)$



Instance of \rightarrow^+

newtype $a \rightarrow^+ b = \text{AddFun } (a \rightarrow b)$

instance *Category* (\rightarrow^+) **where**

type *Obj* $(\rightarrow^+) = \text{Additive}$

id = *AddFun id*

AddFun g \circ *AddFun f* = *AddFun* (*g* \circ *f*)

instance *Monoidal* (\rightarrow^+) **where**

AddFun f \times *AddFun g* = *AddFun* (*f* \times *g*)

instance *Cartesian* (\rightarrow^+) **where**

exl = *AddFun exl*

exr = *AddFun exr*

dup = *AddFun dup*

Instance of \rightarrow^+

instance *Cocartesian* (\rightarrow^+) **where**

inl = *AddFun inlF*

inr = *AddFun inrF*

jam = *AddFun jamF*

inlF :: *Additive* $b \Rightarrow a \rightarrow a \times b$

inrF :: *Additive* $a \Rightarrow b \rightarrow a \times b$

jamF :: *Additive* $a \Rightarrow a \times a \rightarrow a$

inlF = $\lambda a \rightarrow (a, 0)$

inrF = $\lambda b \rightarrow (0, b)$

jamF = $\lambda(a, b) \rightarrow a + b$

NumCat definition

class *NumCat* *k a* **where**

negateC :: *a* ' *k* ' *a*

addC :: (*a* × *a*) ' *k* ' *a*

mulC :: (*a* × *a*) ' *k* ' *a*

...

instance *Num* *a* ⇒ *NumCat* (→) *a* **where**

negateC = *negate*

addC = *uncurry* (+)

mulC = *uncurry* (*)

...

$$\mathcal{D}(\text{negate } u) = \text{negate } (\mathcal{D} u)$$

$$\mathcal{D}(u + v) = \mathcal{D} u + \mathcal{D} v$$

$$\mathcal{D}(u * v) = u * \mathcal{D} v + v * \mathcal{D} u$$

- Imprecise on the nature of u and v .
- A precise and simpler definition would be to differentiate the operations themselves.

class *Scalable* *k a* **where**

scale :: $a \rightarrow (a \text{ ' } k \text{ ' } a)$

instance *Num* *a* \Rightarrow *Scalable* (\rightarrow^+) *a* **where**

scale *a* = *AddFun* $(\lambda da \rightarrow a * da)$

instance *NumCat* *D* **where**

negateC = *linearD* *negateC*

addC = *linearD* *addC*

mulC = *D* $(\lambda(a, b) \rightarrow (a * b, \text{scale } b \nabla \text{scale } a))$

instance *FloatingCat* *D* **where**

sinC = *D* $(\lambda a \rightarrow (\sin a, \text{scale } (\cos a)))$

cosC = *D* $(\lambda a \rightarrow (\cos a, \text{scale } (-\sin a)))$

expC = *D* $(\lambda a \rightarrow \text{let } e = \exp a \text{ in } (e, \text{scale } e))$

...

Examples

$sqr :: Num\ a \Rightarrow a \rightarrow a$

$sqr\ a = a * a$

$magSqr :: Num\ a \Rightarrow a \times a \rightarrow a$

$magSqr\ (a, b) = sqr\ a + sqr\ b$

$cosSinProd :: Floating\ a \Rightarrow a \times a \rightarrow a \times a$

$cosSinProd\ (x, y) = (cos\ z, sin\ z) \textbf{ where } z = x * y$

With a compiler plugin we can obtain

$sqr = mulC \circ (id \Delta id)$

$magSqr = addC \circ (mulC \circ (exl \Delta exl) \Delta mulC \circ (exr \Delta exr))$

$cosSinProd = (cosC \Delta sinC) \circ mulC$

Generalizing Automatic Differentiation

newtype D_k a $b = D (a \rightarrow b \times (a ' k ' b))$

linearD :: $(a \rightarrow b) \rightarrow (a ' k ' b) \rightarrow D_k a b$

linearD f $f' = D (\lambda a \rightarrow (f a, f'))$

instance *Category* $k \Rightarrow$ *Category* D_k **where**

type *Obj* $D_k =$ *Additive* \wedge *Obj* k ...

instance *Monoidal* $k \Rightarrow$ *Monoidal* D_k **where** ...

instance *Cartesian* $k \Rightarrow$ *Cartesian* D_k **where** ...

instance *Cocartesian* $k \Rightarrow$ *Cocartesian* D_k **where**

inl = *linearD* *inlF* *inl*

inr = *linearD* *inrF* *inr*

jam = *linearD* *jamF* *jam*

instance *Scalable* $k\ s \Rightarrow \text{NumCat } D_k\ s$ **where**
negateC = *linearD negateC negateC*
addC = *linearD addC addC*
mulC = *D* ($\lambda(a, b) \rightarrow (a * b, \text{scale } b \nabla \text{scale } a)$)

Matrices

There exists three, non-exclusive, possibilities for a nonempty matrix W :

- width $W = \text{height } W = 1$;
- W is the horizontal juxtaposition of two matrices U and V , where height $W = \text{height } U = \text{height } V$ and width $W = \text{width } U + \text{width } V$;
- W is the vertical juxtaposition of two matrices U and V , where width $W = \text{width } U = \text{width } V$ and height $W = \text{height } U + \text{height } V$.

Extracting a Data Representation

In machine learning, a Gradient-based optimization works by searching for local minima in the domain of a differentiable function $f :: a \rightarrow s$. Each step in the search is in the direction opposite of the gradient of f , which is a vector form of $\mathcal{D} f$.

Given a linear map $f' :: U \multimap V$ represented as a function, it is possible to extract a Jacobian matrix by applying f to every vector in a basis of U .

Extracting a Data Representation

In machine learning, a Gradient-based optimization works by searching for local minima in the domain of a differentiable function $f :: a \rightarrow s$. Each step in the search is in the direction opposite of the gradient of f , which is a vector form of $\mathcal{D} f$.

Given a linear map $f' :: U \multimap V$ represented as a function, it is possible to extract a Jacobian matrix by applying f to every vector in a basis of U .

Generalized Matrices

Given a scalar field s , a free vector space has the form $p \rightarrow s$ for some p , where the cardinality of p is the dimension of the vector space and there exists a finite number of values for p .

In particular, we can represent vector spaces over a given field as a representable functor, i.e., a functor F such that:

$$\exists p \forall s \ F \ s \cong \ p \rightarrow s$$

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A short introduction

- We've derived and generalized an AD algorithm using categories.
- With fully right-associated compositions this algorithm becomes a forward-mode AD and with fully left-associated becomes a reverse-mode AD.
- We want to obtain generalized FAD and RAD algorithms.
- How do we describe this in Categorical notation?

Converting morfisms

Given a category k we can represent its morfisms the following way:

Left-Compose functions

$f :: a' \rightarrow k' \rightarrow b \Rightarrow (\circ f) :: (b' \rightarrow k' \rightarrow r) \rightarrow (a' \rightarrow k' \rightarrow r)$ where r is any object of k .

If h is the morfism we'll compose with f then h is the continuation of f .

Instance deduction

Defining new type

newtype $\text{Cont}_k^f a b = \text{Cont} ((b \text{ ' } k' \text{ } r) \rightarrow (a \text{ ' } k' \text{ } r))$

Functor derived from type

$\text{cont} :: \text{Category } k \Rightarrow (a \text{ ' } k' \text{ } b) \rightarrow \text{Cont}_k^f a b$
 $\text{cont } f = \text{Cont } (\circ f)$

Instance deduction

instance *Category* $k \Rightarrow \text{Category } \text{Cont}_k^r$ **where**

id = *Cont id*

Cont g \circ *Cont f* = *Cont (f* \circ *g)*

instance *Monoidal* $k \Rightarrow \text{Monoidal } \text{Cont}_k^r$ **where**

Conf f \times *Cont g* = *Cont (join* \circ (*f* \times *g*) \circ *unjoin)*

instance *Cartesian* $k \Rightarrow \text{Cartesian } \text{Cont}_k^r$ **where**

exl = *Cont (join* \circ *inl)*; *exr* = *Cont (join* \circ *inr)*

dup = *Cont (jam* \circ *unjoin)*

instance *Cocartesian* $k \Rightarrow \text{Cocartesian } \text{Cont}_k^r$ **where**

inl = *Cont (exl* \circ *unjoin)*; *inr* = *Cont (exr* \circ *unjoin)*

jam = *Cont (join* \circ *dup)*

instance *Scalable* k $a \Rightarrow \text{Scalable } \text{Cont}_k^r$ a **where**

scale s = *Cont (scale s)*

A short introduction

Due to its widespread use in ML we'll talk about a specific case of RAD: computing gradients (derivatives of functions with scalar codomains).

A vector space A over a scalar field s has $A \multimap s$ as its dual. Each linear map in $A \multimap s$ can be represented in the form of $\text{dot } u$ for some $u :: A$ where

Definition and instanciation

class *HasDot* (S) u **where** $\text{dot} :: u \rightarrow (u \multimap s)$

instance *HasDot* (IR) IR **where** $\text{dot} = \text{scale}$

instance (*HasDot* (S) a , *HasDot* (S) b)

\Rightarrow *HasDot* (S) ($a \times b$)

where $\text{dot } (u, v) = \text{dot } u \Delta \text{dot } v$

Instance deduction

The internal representation of $Cont^s_{\circ} a b$ is $(b \multimap s) \rightarrow (a \multimap s)$ which is isomorphic to $(a \rightarrow b)$.

Type definition for duality

`newtype` $Dual_k a b = Dual (b 'k' a)$

Instance deduction

All we need to do to create dual representations of linear maps is to convert from $Cont_k^S$ to $Dual_k$ using a functor:

Functor definition

$$\begin{aligned} asDual &:: (HasDot\ (S)\ a, HasDot\ (S)\ b) \Rightarrow \\ &\quad ((b \multimap s) \rightarrow (a \multimap s)) \rightarrow (b \multimap a) \\ asDual\ (Cont\ f) &= Dual\ (onDot\ f) \end{aligned}$$

where

$$\begin{aligned} onDot &:: (HasDot\ (S)\ a, HasDot\ (S)\ b) \Rightarrow \\ &\quad ((b \multimap s) \rightarrow (a \multimap s)) \rightarrow (b \multimap a) \\ onDot\ f &= dot^{-1} \circ f \circ dot \end{aligned}$$

Instance deduction

instance *Category* $k \Rightarrow \text{Category } \text{Dual}_k$ **where**

id = *Dual id*

Dual g \circ *Dual f* = *Dual (f* \circ *g)*

instance *Monoidal* $k \Rightarrow \text{Monoidal } \text{Dual}_k$ **where**

Dual f \times *Dual g* = *Dual (f* \times *g)*

instance *Cartesian* $k \Rightarrow \text{Cartesian } \text{Dual}_k$ **where**

exl = *Dual inl*; *exr* = *Dual inr*

dup = *Dual jam*

instance *Cocartesian* $k \Rightarrow \text{Cocartesian } \text{Dual}_k$ **where**

inl = *Dual exl*; *inr* = *Dual exr*

jam = *Dual dup*

instance *Scalable* $k \Rightarrow \text{Scalable } \text{Dual}_k$ **where**

scale s = *Dual (scale s)*

Final notes

- (∇) and (Δ) mutually dualize
 $(Dual\ f\ \nabla\ Dual\ g) = Dual\ (f\ \Delta\ g)$ and $Dual\ f\ \Delta\ Dual\ g = Dual\ (f\ \nabla\ g)$
- Using the definition from chapter 8 we can determine that the duality of a matrix corresponds to its transposition

Fowards-mode Automatic Differentiation(FAD)

We can use the same deductions we've done in Cont and Dual to derive a category with full right-side association, thus creating a generized FAD algorithm.

This algorithm is far more appropriate for low dimension domains.

Type definition and functor from type

newtype $Begin_k^r a b = Begin ((r ' k' a) \rightarrow (r ' k' b))$

$begin :: Category k \Rightarrow (a ' k' b) \rightarrow Begin_k^r a b$

$begin f = Begin (f \circ)$

We can derive categorical instances from the functor above and we can choose r to be the scalar field s , noting that $s \multimap a$ is isomorphic to a .

Scaling Up

- Practical applications often involve high-dimensional spaces.
- Binary products are a very inefficient and unwieldy way of encoding high-dimensional spaces.
- A practical alternative is to consider n-ary products as representable functors.

class *Category* $k \Rightarrow \text{Monoidall } k \ h$ **where**

crossl :: $h \ (a' \ k' \ b) \rightarrow (h \ a' \ k' \ h \ b)$

instance *Zip* $h \Rightarrow \text{Monoidall } (\rightarrow) \ h$ **where**

crossl = *zipWith id*

class *Monoidall* $k\ h \Rightarrow \text{CartesianI } k\ h$ **where**

exl $:: h\ (h\ a\ 'k\ 'a)$

repl $:: a\ 'k\ 'h\ a$

instance (*Representable* h , *Zip* h , *Pointed* h) \Rightarrow

CartesianI $(\rightarrow)\ h$ **where**

exl = *tabulate* (*flip index*)

repl = *point*

- The following is not the class the author was thinking

class *Representable* h **where**

type *Rep* $h :: *$

tabulate $:: (Rep\ h \rightarrow a) \rightarrow h\ a$

index $:: h\ a \rightarrow Rep\ h \rightarrow a$

class *Monoidall* *k h* \Rightarrow *Cocartesianl* *k h* **where**

inl :: *h* (*a* ' *k* ' *h a*)

jaml :: *h a* ' *k* ' *a*

instance (*Monoidall* *k h*, *Zip h*) \Rightarrow *Monoidall* *D_k h* **where**

crossl *fs* = *D* ((*id* \times *crossl*) \circ *unzip* \circ *crossl* (*fmap unD fs*))

instance (*Cocartesianl* (\rightarrow) *h*, *Cartesianl* *k h*, *Zip h*) \Rightarrow

Cartesianl *D_k h* **where**

exl = *linearD* *exl* *exl*

repl = *zipWith linearD repl repl*

instance (*Cocartesianl* *k h*, *Zip h*) \Rightarrow *Cocartesianl* *D_k h* **where**

inl = *zipWith linearD inlF inl*

jaml = *linearD sum jaml*

class *Monoidall* *k h* \Rightarrow *Cocartesianl* *k h* **where**

inl :: *h* (*a* ' *k* ' *h a*)

jaml :: *h a* ' *k* ' *a*

instance (*Monoidall* *k h*, *Zip h*) \Rightarrow *Monoidall* *D_k h* **where**

crossl *fs* = *D* ((*id* \times *crossl*) \circ *unzip* \circ *crossl* (*fmap unD fs*))

instance (*Cocartesianl* (\rightarrow) *h*, *Cartesianl* *k h*, *Zip h*) \Rightarrow

Cartesianl *D_k h* **where**

exl = *zipWith linearD* *exl* *exl*

repl = *linearD* *repl* *repl*

instance (*Cocartesianl* *k h*, *Zip h*) \Rightarrow *Cocartesianl* *D_k h* **where**

inl = *zipWith linearD inlF* *inl*

jaml = *linearD sum* *jaml*

Conclusion

- Suggests that some of the work referred to does just a part of this article.
- This article ([Elliott 2018][2]) is a follow up of [Elliott 2017][1]
- Suggests that this implementation is simple, efficient, it can free memory dinamically (RAD) and is naturally parallel.
- Future work are detailed performace analysis; higher-order differentiation and automatic incrementation (continuing previous work [Elliott 2017][1])



ELLIOTT, C.

Compiling to categories.

Proc. ACM Program. Lang. 1, ICFP (Aug. 2017),
27:1–27:27.



ELLIOTT, C.

The simple essence of automatic differentiation.

Proc. ACM Program. Lang. 2, ICFP (July 2018),
70:1–70:29.