

Functional Differentiation of Computer Programs

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Outline

Motivation & Introduction

Differentiation techniques

1st approach

Final approach

Applications

Conclusion

References

Why do we want to compute derivatives ?

Derivatives are useful for ...

- solving **Optimization Problems**
- **Image Processing** (Feature Extraction, Object Recognition)
- **3-D-Modelling** (geom. properties of curves and surfaces)
- Many fields of **scientific computing** like engineering, ...

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We show a

- **purely functional** implementation (using **Haskell**)
- only based on **numerics** (no symbolic computations)
- relying on **overloading** of arithmetic operators, **lazy evaluation** and **type classes** concept
- yielding (point-wise) derivatives of ..
- .. *any* order, using '*co-recursive*' data structures and
- .. *any* mathematical function definable in Haskell code

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3 ways ... (I)

We have 3 ways to compute derivatives:

1. Finite differences approximation:

$$f'(x) \approx \frac{f(x + \Delta x) - f(x)}{\Delta x}$$

- *Inaccurate* if Δx is too big,
- *Cancellation errors* if Δx is too small.

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2. Symbolic differentiation: 'manual', formal method

- Exact, but *quite costly*
- Control structures like loops, etc. have to be 'unfolded' \rightsquigarrow symbolic interpretation of whole program

3 ways ... (II)

3. Computational Differentiation - **CD**: Our approach !

- **Numeric** algorithms, based on standard arithmetic operations, with known differential properties (school knowledge!)
- As *exact* as numerical evaluation of symbolic derivatives (but lacks symbolical (analytical) results)
based on **overloading** (already implemented in C++)
- Functional implementation relies on **co-recursive data structures**

$$R \alpha = C \alpha \mid T \alpha (R \alpha)$$

for computing derivatives of *any* order!

- **Drawback**: discontinuous or non-differentiable functions (e.g. `abs x`) also yield values for their derivatives, which is unsatisfactory

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First approach: 'We are not lazy!'

We start with a simple approach

- only compute **first derivatives**
- *without* lazy evaluation
- yielding a quite efficient solution
- introduce **'extended numerical' structure:**

```
type Dx = (Double, Double)
```

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- only compute **first derivatives**
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- introduce **'extended numerical' structure:**

type Dx = (Double, Double)

- grouping numerical value (**main value**) e of an expression with value of first derivative e' *at the same point*: (e, e')
- $(c, 0.0)$ for constants c and $(x, 1.0)$ for variables x .
- Could replace `double` by any ring $(R, +, \times)$ or field $(F, +, \times, /)$
- Remark: No symbolic calculations \rightsquigarrow constants and variables don't need to have explicit names!
e.g.: $(3.141, 0.0)$ or $(2.523, 1.0)$

Overloaded Arithmetic

- Define overloaded arithmetic operators for type $\mathbb{D}x$
- implementing basic derivation laws
sum-, product-, quotient-rule, ...

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$$(x, a) + (y, b) = (x + y, a + b) \quad (::: Dx \rightarrow Dx \rightarrow Dx)$$

$$(x, a) - (y, b) = (x - y, a - b)$$

$$(x, a) * (y, b) = (x * y, x * b + a * y)$$

$$\text{negate } (x, a) = (\text{negate } x, \text{negate } a)$$

$$(x, a) / (y, b) = (x / y, (a * y - x * b) / (y * y))$$

$$\text{recip } (x, a) = (w, (\text{negate } a) * w * w) \text{ where } w = \text{recip } x$$

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- Also auxiliary functions to construct constants and variables and a conversion function

$$\text{dCst } z = (z, 0.0) \quad \text{dVar } z = (z, 1.0)$$

$$\text{fromDouble } z = \text{dCst } z$$

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- **Chain rule:** $d(f(g(x))) = f'(g(x)) \cdot d(g(x))$
- Important for derivatives of elementary functions like \sin, \cos, \log, \dots
- These functions f are **lifted** to the D_x domain, given their *derivative form* f'

`dlift f f' (x,a) = (f x , a * f' x)`

`exp = dlift exp exp`

`sin = dlift sin cos`

- .. same for \cos, \sqrt{x}, \log
- Now we can define arbitrary complicated mathematical functions like $f\ x = x*x * \cos(x)$
- .. and $f\ 6.5 \rightsquigarrow (41.260827, 3.606820) \equiv (f(6.5), f'(6.5))$

Haskell type classes

- Approach doesn't use Haskell's type classes ¹
- Introduce modified **algebraic style** library (\equiv mathematical hierarchy) of type classes:

¹generic operations: declared within classes, datatypes accepting them are instances of them

Haskell type classes

- Approach doesn't use Haskell's type classes ¹
- Introduce modified **algebraic style** library (\equiv mathematical hierarchy) of type classes:
 - `AddGroup` for addition and subtraction
 - `Monoid` for multiplication, `Group` for division
 - `Ring` for *structures* supporting addition and multiplication, `Field` adding division
 - `Module` abstracts multiplication of complex object by element of basic domain (e.g.: $\lambda \cdot \vec{v}$)
 - `Number` uses `fromInt`, `fromDouble` to convert standard numbers in our `Dx` domain

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Differential Algebra and 'Lazy towers of derivatives'

- Compute (as promised) 'all' derivatives of functions (exact: **an a priori unknown number**)
- Data structure, representing expression of **infinite domain**: num. value e_0 and all derivatives $[e_0, e_1, e_2, \dots]$ ($e_i \equiv e^{(i)}$) **without explicit truncation, created by co-recursion!**

Differential Algebra and 'Lazy towers of derivatives'

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- Data structure, representing expression of **infinite domain**: num. value e_0 and all derivatives $[e_0, e_1, e_2, \dots]$ ($e_i \equiv e^{(i)}$) **without explicit truncation, created by co-recursion!**
- Need background in **Differential Algebra**
- Field $(F, +, \times, /)$ with derivation $a \mapsto a'$
- $F = \mathbb{R}$ is trivial: $\forall x \in \mathbb{R} : x \mapsto 0$
- Extend field to $F(x)$ by adjoining symbolic x
- If mathematical structure of the expressions known, we can discard the $x \rightsquigarrow$ no symbolic computations
- E.g.: Represent polynomial by list of its coefficients

Get it started

- **Important:** We assume that x and x' are algebraic independent and thus assign to expressions e all derivatives e', e'', \dots by the derivation operator $e_n \mapsto e_{n+1}$
- We use no indeterminate and just operate on **infinite, lazy lists** of a priori independent elements
- We define the **co-recursive**, infinite, parameterized type

```
data Dif a = C a | D a (Dif a)
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- $C\ a$ codes a constant a whose derivative is 0
- $D\ e\ (D\ a\ (D\ b\ \dots))$ codes the numerical value of the expression (e) and the remainder the **tower of derivatives** ($a = e', b = e'', \dots$)
- In general, a should be an instance of a field, e.g. `Double`

Overloaded Arithmetics for Dif domain

- The **derivation operator** `df :: a -> a` is declared in `class Dif a`
- Lifting* procedures: `df (C _) = C 0.0 ; df (D _ p) = p`
- We implement the basic **derivation laws**
- The **sum-rule** is trivial, with `Dif a` instance of `AddGroup` class:

$$C\ x + C\ y = C\ (x+y)$$

$$C\ x + D\ y\ y' = D\ (x+y)\ y'$$

$$D\ x\ x' + D\ y\ y' = D\ (x+y)\ (x'+y')$$

$$\text{neg} = \text{fmap}\ \text{neg}$$

$$^2x^* > s = \text{fmap}\ (x^*)\ s$$

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- Same for **product-rule** and **unaltered constants** (`Monoid` class):

$$C\ x * C\ y = C\ (x*y)$$

$$C\ x * p = x*>p$$

$$p@(D\ x\ x') * q@(D\ y\ y') = D\ (x*y)\ (x'*q+p*y')$$

$$^2x*>s = \text{fmap}\ (x*)\ s$$

Overloaded Arithmetics (II)

- **Reciprocal** $(\frac{1}{u(x)})' = \frac{-u'(x)}{u(x)^2}$ heavily uses *lazy evaluation*
(Group class):

```
recip (C x) = C (recip x)
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```
recip (D x x') = ip where
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ip = D (recip x) (neg x'*ip*ip)
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- further trivial cases left out !

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- further trivial cases left out !
- **Division** might present some problems: $\frac{0}{0}$

```
p@(D x x') / q@(D y y')
```

```
| x==0.0 && y==0.0 = x'/y' --L' Hopital--
```

```
| otherwise = D (x/y) (x'*q - p*y'/(q*q))
```

Lifting and the chain rule

- **Transcendental functions** f like \exp, \sin, \dots need *lifting* to the `Dif` domain
- Definition of their **list of formal derivatives** f_q , using *lazy evaluation* (`Group` class)
- E.g.: $(\exp(u(x)))' = u'(x) \cdot \exp(u(x))$

```
dlift (f:fq) p@(D x x') =
  D (f x) (x' * dlift fq p) {--Chain rule--}
```

```
exp (D x x') = r where r = D (exp x) (x'*r)
```

```
sin =
  dlift (cycle[sin,cos,(neg . sin),(neg . cos)])
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- \cos, \log, \sqrt{x} in the same manner!
- and that's it ... **we're done !!!**
- Now: $df (df (df (f 6.5))) \rightsquigarrow -30.288818 \equiv f'''(6.5)$

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Example applications

- Wide spread, huge application domain, 'ranging from reactor diagnostic, meteorology, oceanography, up to biostatistics' and quantum theory

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- **One example:** Elegant coding of differential recurrences, like the *Hermite function*, without explicit truncation of **recurrent computation** !

$$H_0(x) = \exp\left(\frac{-x^2}{2}\right)$$

$$H_n(x) = \frac{1}{\sqrt{2n}} \left(x \cdot H_{n-1}(x) - \frac{d}{dx} (H_{n-1}(x)) \right)$$

herm n x = cc where

D cc _ = hr n (dVar x)

hr 0 x = exp(neg x * x / fromDouble 2.0)

hr n x = (x*z - df z)/(sqrt(fromInteger (2*n)))

where z=hr (n-1) x

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Final Remarks - Pro's and Con's

- **clear, readable, compact** (especially for towers!) and **semantically powerful** \rightsquigarrow **nice coding tool!**
- *Thunks* of lazy evaluation may introduce **space leaks**, when computing derivatives of high order
Remedy: use truncated strict variant, like 1st approach, given number of derivatives to compute
- not extremely efficient, hence outperformed by C++ implementations and semi-automatic systems
- Still **useable** and faster than symbolic systems

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Remedy: use truncated strict variant, like 1st approach, given number of derivatives to compute
- not extremely efficient, hence outperformed by C++ implementations and semi-automatic systems
- Still **useable** and faster than symbolic systems
- *Claim*: straight forward **generalization** to vector or tensor objects
- Control structures (*if-then-else*) need **arithm. relations** on (infinite) `Dif` type
Simplified **remedy**: just compare main values

Summary

- We've seen: Rewarding application of **modern functional programming paradigms** to scientific computing (usually domain of low-level languages)

Contribution

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Contribution

- Type inference, Overloading \Rightarrow **overloaded arithmetic operators**, declare differentiation *variables*
- Lazy evaluation \Rightarrow **derivation operator**, applicable arbitrary (*a priori* unknown) number of times, without explicit truncation!
- Type classes, Lifting \Rightarrow **extended arithmetics**, valid for any *basic domain*, e.g.: \mathbb{C} , \mathbb{P}

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