

Algorithmic Differentiation I

First Directional Derivatives of Univariate Scalar Functions

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Informatik 12:
Software and Tools for Computational Engineering (STCE)

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Objective and Learning Outcomes

Recall

First-Order Tangents

Finite Difference Approximation
dco/c++

Second-Order Tangents

Finite Difference Approximation
dco/c++

Higher-Order Tangents

Summary and Next Steps

Objective and Learning Outcomes

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Summary and Next Steps

Objective

- ▶ Introduction to first- and second-order algorithmic differentiation of univariate scalar functions with `dco/c++`

Learning Outcomes

- ▶ You will understand
 - ▶ tangent-mode algorithmic differentiation
 - ▶ finite difference approximation of directional derivatives of univariate scalar functions.
- ▶ You will be able to
 - ▶ use `dco/c++` for the computation of derivatives of arbitrary order for implementation of univariate scalar functions as C++ programs
 - ▶ compare the results with approximations computed by finite differences

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Summary and Next Steps

Let $f : \mathbf{R} \rightarrow \mathbf{R}$ be n -times continuously differentiable.

Given the value of $f(x)$ at some point $\tilde{x} \in \mathbf{R}$ the function value $f(\tilde{x} + \Delta x)$ at a neighboring point can be approximated by a **Taylor series** as

$$f(\tilde{x} + \Delta x) \approx_{O(\Delta x^n)} f(\tilde{x}) + \sum_{k=1}^{n-1} \frac{1}{k!} \cdot \frac{d^k f}{dx^k}(\tilde{x}) \cdot \Delta x^k .$$

Throughout this course we assume convergence of the Taylor series for $k \rightarrow \infty$ to the true value of $f(\tilde{x} + \Delta x)$ within all subdomains of interest, which is not the case for arbitrary functions.

For $n = 4$ we get

$$f(\tilde{x} + \Delta x) = f(\tilde{x}) + f'(\tilde{x}) \cdot \Delta x + \frac{1}{2} \cdot f''(\tilde{x}) \cdot \Delta x^2 + \frac{1}{6} \cdot f'''(\tilde{x}) \cdot \Delta x^3 + O(|\Delta x|^4) .$$

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Summary and Next Steps

First-order tangents

$$f^{(1)} : \mathbf{R} \times \mathbf{R} \rightarrow \mathbf{R} : y^{(1)} = f^{(1)}(x, x^{(1)})$$

of

$$f : \mathbf{R} \rightarrow \mathbf{R} : y = f(x)$$

are **first directional derivatives**, that is, they are scaled first derivatives of f , i.e.,

$$y^{(1)} = f^{(1)}(x, x^{(1)}) \equiv f'(x) \cdot x^{(1)},$$

where

$$f'(x) \equiv \frac{df}{dx}(x) \in \mathbf{R}.$$

Superscripts $*^{(1)}$ are used to denote first-order tangents.

The solution of linear equations amounts to simple scalar division. The solution of nonlinear equations can be challenging.



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Many numerical methods for nonlinear problems are built on local (at \tilde{x}) replacement of the target function with a **linear** (affine; in Δx) **approximation** derived from the truncated Taylor series expansion and “hoping” that

$$f(\tilde{x} + \Delta x) \approx f(\tilde{x}) + f'(\tilde{x}) \cdot \Delta x ,$$

i.e, hoping for a reasonably small remainder.

The solution of a sequence of linear problems is then expected to yield an iterative approximation of the solution to the nonlinear problem.

Note that the **linearization** of a nonlinear function f at some point \tilde{x} is a function in Δx :

$$\bar{f}(\Delta x) = f(\tilde{x}) + f'(\tilde{x}) \cdot \Delta x .$$

Under the assumption that

$$f(\tilde{x} + \Delta x) \approx f(\tilde{x}) + f'(\tilde{x}) \cdot \Delta x$$

the derivative of $\bar{f}(\Delta x)$ wrt. Δx

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

can be used to approximate the derivative $f'(\tilde{x})$ of the nonlinear function f .
This method is known as **finite difference** approximation.

Finite differences can be used to approximate tangents at a given point $\tilde{x} \in \mathbf{R}$.

$$f'(\tilde{x}) \cdot x^{(1)} \approx_1 \frac{f(\tilde{x} + \Delta x \cdot x^{(1)}) - f(\tilde{x})}{\Delta x} \quad (\text{forward})$$

$$\approx_1 \frac{f(\tilde{x}) - f(\tilde{x} - \Delta x \cdot x^{(1)})}{\Delta x} \quad (\text{backward})$$

$$\approx_2 \frac{f(\tilde{x} + \Delta x \cdot x^{(1)}) - f(\tilde{x} - \Delta x \cdot x^{(1)})}{2 \cdot \Delta x} \quad (\text{central})$$

where $\Delta x = \Delta x(\tilde{x}) \in \mathbf{R}$ is a suitable perturbation.

Forward and backward finite differences exhibit first-order accuracy (\approx_1 ; error scales with Δx) while central finite differences turn out to be second-order accurate (\approx_2 ; error scales with Δx^2).

dco/c++ (derivative code by overloading in C++)¹

- ▶ is an algorithmic differentiation tools developed by STCE in collaboration with the Numerical Algorithms Group Ltd. (nag.co.uk)
- ▶ supports the (semi-)automatic generation of **derivative code of arbitrary order** for numerical simulations written in C++.
- ▶ uses **overloading** of arithmetic operators (e.g, +, \) and elementary functions (e.g, sin, log) for a set of custom data types (e.g, **gt1s**) to implement algorithmic differentiation.
- ▶ is actively used by numerical simulation projects world-wide.

¹Lotz, Leppkes, Naumann: dco/c++: Derivative Code by Overloading in C++. Aachener Informatik-Berichte AIB-2011-06.

```
1  template<typename T> void f(const T& x, T& y);
2
3  template<typename T>
4  void f_t(const T& x_v, T& y_v, T& dydx) {
5      using DCO_T=typename dco::gt1s<T>::type;
6      DCO_T x,y;
7      dco::value(x)=x_v; dco::derivative(x)=1;
8      f(x,y);
9      y_v=dco::value(y); dydx=dco::derivative(y);
10 }
11
12 int main() {
13     double x=1, y, dydx;
14     f_t(x,y,dydx); ...
15     return 0;
16 }
```

```
1  template<typename T> void f(const T& x, T& y);
2
3  template<typename T>
4  void f_cfd(const T& x, T& y, T& dydx) {
5      f(x,y);
6      T dx=fabs(x)<1 ? sqrt(std::numeric_limits<T>::epsilon())
7          : sqrt(std::numeric_limits<T>::epsilon())*fabs(x);
8      T yp,ym;
9      f(x+dx,yp); f(x-dx,ym);
10     dydx=(yp-ym)/(2*dx);
11 }
12
13 int main() {
14     double x=1, y, dydx;
15     f_cfd(x,y,dydx); ...
16     return 0;
17 }
```

Objective and Learning Outcomes

Recall

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Higher-Order Tangents

Summary and Next Steps

Second-order tangents

$$f^{(1,2)} : \mathbb{R} \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R} : y^{(1,2)} = f^{(1,2)}(x, x^{(1)}, x^{(2)})$$

of

$$f : \mathbb{R} \rightarrow \mathbb{R} : y = f(x)$$

are first **directional derivatives of $f^{(1)}$** in direction $x^{(2)}$, that is, they are scaled first derivatives of $f^{(1)}$ wrt. x , i.e.,

$$y^{(1,2)} = f^{(1,2)}(x, x^{(1)}, x^{(2)}) \equiv f''(x) \cdot x^{(1)} \cdot x^{(2)},$$

where

$$f''(x) \equiv \frac{df'}{dx}(x) = \frac{d^2f}{dx^2}(x) \in \mathbb{R}.$$

Superscripts $*$ ⁽²⁾ are used to denote first-order tangents of $f^{(1)}$.

Second-order tangents carry superscripts $*$ ^(1,2) \equiv $*$ ⁽¹⁾⁽²⁾.

Second derivatives can be approximated as derivatives of [a finite difference approximation of] f' .

$$\begin{aligned} f''(x) &\approx \frac{f'(x + \Delta x, \Delta x) - f'(x - \Delta x, \Delta x)}{2 \cdot \Delta x} \\ &= \frac{f(x + 2 \cdot \Delta x) - 2 \cdot f(x) + f(x - 2 \cdot \Delta x)}{4 \cdot \Delta x^2} \end{aligned}$$

The first expression yields a natural approach to implementing second-order finite differences.

```
1  template<typename T> void f(const T& x, T& y);
2
3  template<typename T> void f_cfd(const T& x, T& y, T& dydx);
4
5  template<typename T>
6  void f_cfd_cfd(const T& x, T& y, T& dydx, T& ddydxx) {
7      f_cfd(x,y,dydx);
8      T dx=fabs(x)<1 ? sqrt(sqrt(std::numeric_limits<T>::epsilon()))
9          : sqrt(sqrt(std::numeric_limits<T>::epsilon()))*fabs(x);
10     T dummy,dydyp,dydym;
11     f_cfd(x+dx,dummy,dydyp); f_cfd(x-dx,dummy,dydym);
12     ddydxx=(dydyp-dydym)/(2*dx);
13 }
14
15 int main() {
16     double x=1, y, dydx, ddydxx;
17     f_cfd_cfd(x,y,dydx,ddydxx);
18     // ...
19     return 0;
20 }
```

Investigate **numerical effects** due to finite precision floating-point arithmetic!

```
1  template<typename T> void f(const T& x, T& y);
2
3  template<typename T> void f_t(const T& x_v, T& y_v, T& dydx);
4
5  template<typename T>
6  void f_t_t(const T& x_v, T& y_v, T& dydx_v, T& ddydxx) {
7      using DCO_T=typename dco::gt1s<T>::type;
8      DCO_T x,y,dydx;
9      dco::value(x)=x_v; dco::derivative(x)=1;
10     f_t(x,y,dydx);
11     y_v=dco::value(y);
12     dydx_v=dco::value(dydx);
13     ddydxx=dco::derivative(dydx);
14 }
15
16 int main() {
17     double x=1, y, dydx, ddydxx;
18     f_t_t(x,y,dydx,ddydxx);
19     // ...
20     return 0;
21 }
```

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The above generalizes to third- and higher-order tangents:

$$\begin{aligned}y^{(1,2,3)} &= f^{(1,2,3)}(x, x^{(1)}, x^{(2)}, x^{(3)}) \\ &\equiv f'''(x) \cdot x^{(1)} \cdot x^{(2)} \cdot x^{(3)} \\ &\dots\end{aligned}$$

where

$$f'''(x) \equiv \frac{df''}{dx}(x) = \frac{d^3f}{dx^3}(x) \in \mathbf{R} \quad \text{etc.}$$

dco/c++ supports tangents of arbitrary order through nesting of its base-type-generic first-order tangent type `gt1s<BASE_TYPE>`.

Finite differences for tangents of third and higher order can rarely be trusted because of numerical issues due to finite precision floating-point arithmetic.

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Summary and Next Steps

Summary

- ▶ Finite differences for tangents of arbitrary order suffers from numerical issues due to finite precision floating-point arithmetic.
- ▶ dco/c++ delivers tangents of arbitrary order with machine accuracy.

Next Steps

- ▶ Run sample code and compare results.
- ▶ Check out dco/c++ for higher derivatives.
- ▶ Continue the course to find out more ...