

# Numerical Differentiation

## A Short Introduction



Hans W. Borchers  
ABB Corporate Research



# Numerical Differentiation / Derivatives

- **Content**
  - Basic Definitions and Formulas
  - Error terms and optimal step size
  - Some difference formulas
  - Partial derivatives
  - Higher derivatives
  - Lagrange Interpolation
  - Richardson's extrapolation
  - Complex-step approximate derivatives
  - Example
  
  - Automated differentiation
  - Symbolic differentiation
  - Software

# Numerical Derivatives in General

## ■ What

**Numerical differentiation computes (or estimates) the derivatives or 'slope' of a function by calculating function values at only a set of discrete points.**

## ■ Why

**Derivative in explicit functional form may not be available:**

- error prone differentiation by hand, or functions as variables
- using functions with unknown derivatives, or available only as finite sets of points
- blackbox functions (i.e. engineering procedures)

## ■ When

**The evaluation or approximation of derivatives plays a central part in many applications:**

- design engineering
- simulation of dynamical systems (e.g. weather forecast)
- numerically solving differential equations
- non-linear optimization problems (Kuhn-Tucker conditions, Hessian)

# Example

$$Fh(x) = \int_0^x \sin(t) \sqrt{1 + \sin(t)} dt$$

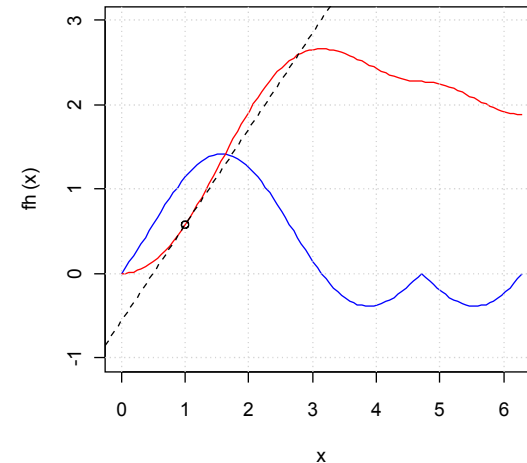
is the antiderivative of

$$fh(x) = \sin(x) \sqrt{1 + \sin(x)}$$

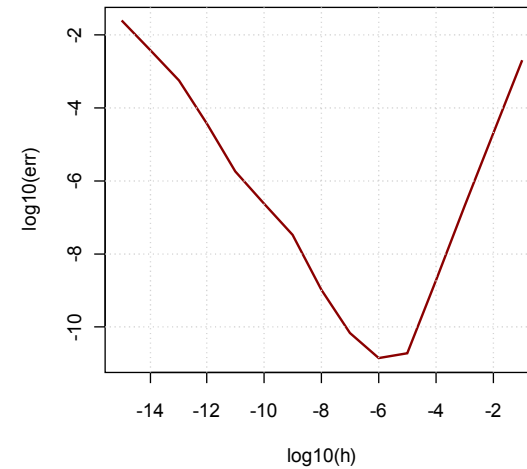
Determine as exactly as possible the derivative at  $x_0 = 1$ .

Method	Value	Difference
True value	1.14188294271546	$1.0 \cdot 10^{-15}$
Forward difference formula	1.14188298816487	$4.5 \cdot 10^{-8}$
Central difference formula	1.14188294264572	$0.7 \cdot 10^{-10}$
Forward three point	1.14188294486617	$2.0 \cdot 10^{-9}$
Central four point	1.14188294246069	$2.5 \cdot 10^{-10}$
Richardson	1.14188294271225	$3.0 \cdot 10^{-12}$
Complex-step		

Functions fh and Fh



Error term



# Basic Formulas and Error terms

- **Forward difference formula**

$$f'(x_0) \approx \frac{f(x_0+h) - f(x_0)}{h}, 0 < h \ll 1$$

- **Central difference formula**

$$f'(x_0) \approx \frac{f(x_0+h) - f(x_0-h)}{2h}, 0 < h \ll 1$$

- **Four-point central difference formula**

$$f'(x_0) \approx \frac{-f(x_0+2h) + 8f(x_0+h) - 8f(x_0-h) + f(x_0-2h)}{12h} + O(h^4),$$
$$0 < h \ll 1$$

- **Optimal step size and accuracy of forward difference formula**

$\varepsilon_m, \varepsilon_f$  machine, function accuracy

$$h \sim \sqrt{\varepsilon_f} \sqrt{\frac{|f|}{|f''|}} \sim \sqrt{\varepsilon_f} \quad \text{step size}$$

$e_t, e_r$  truncation, roundoff error

$$e_t + e_r \sim 2\sqrt{\varepsilon_f} \sqrt{|f| |f''|} \sim \sqrt{\varepsilon_f}$$

- **and central difference formula**

$$h \sim \varepsilon_f^{1/3}$$

$$e_t + e_r \sim \varepsilon_f^{2/3}$$

# Higher Derivatives

## ■ Centered Difference Formulas

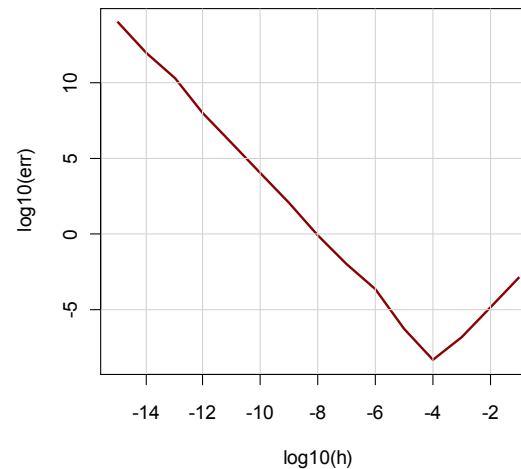
$$f''(x_0) \approx \frac{1}{h^2} [f(x_0+h) - 2f(x_0) + f(x_0-h)]$$

$$f'''(x_0) \approx \frac{1}{2h^3} [f(x_0+2h) - 2f(x_0+h) + 2f(x_0-h) - f(x_0-2h)]$$

$$f^{(4)}(x_0) \approx \frac{1}{h^4} [f(x_0+2h) - 4f(x_0+h) + 6f(x_0) - 4f(x_0-h) + f(x_0-2h)]$$

These formulas are all  $O(h^2)$  and are easily verified by applying the Taylor series.

Error term



$h = 10^{-4}$   
error =  $4.5 \cdot 10^{-9}$

# Partial Derivatives

## ■ Usages

### ■ Partial derivative

$$\frac{\partial f}{\partial x_i} \approx \frac{f(x_1, \dots, x_i+h, \dots, x_n) - f(x_1, \dots, x_i-h, \dots, x_n)}{2h}$$

### ■ Jacobian matrix $\left(\frac{\partial f_i}{\partial x_j}\right)_{1 \leq i \leq m, 1 \leq j \leq m}$

### ■ Laplacian operator $\nabla^2 f = f_{xx} + f_{yy}$

### ■ Biharmonic operator $\nabla^4 f = f_{xxxx} + f_{xyyy} + f_{yyyy}$

### ■ Hessian matrix

$$H(f)_{ij}(x) = \frac{\partial^2 f}{\partial x_i \partial x_j}(x)$$

$$f(x + \Delta x) \approx f(x) + J(x) \Delta x + \frac{1}{2} \Delta x^T H(x) \Delta x$$

## ■ Higher partial derivatives

$$u_{xx}(x_0, y_0) \approx \dot{u}$$

$$\dot{u} \frac{1}{h^2} [u(x_0-h, y_0) - 2u(x_0, y_0) + u(x_0+h, y_0)]$$

$$u_{xy}(x_0, y_0) \approx \frac{1}{4h^2} [-u(x_0-h, y_0+h) + u(x_0+h, y_0+h) + u(x_0-h, y_0-h) - u(x_0+h, y_0-h)]$$

$$\nabla^2 u(x_0, y_0) \approx \frac{1}{h^2} [u(x_0, y_0+h) + u(x_0-h, y_0) - 4u(x_0, y_0) + u(x_0+h, y_0) + u(x_0, y_0-h)]$$

# Discretized Functions

- Lagrange interpolation

$$(x_1, y_1), (x_2, y_2), (x_3, y_3)$$

$$f(x_i) = y_i$$

$$L(x) = L_1(x)y_1 + L_2(x)y_2 + L_3(x)y_3$$

$$L_1(x) = \frac{(x-x_2)(x-x_3)}{(x_1-x_2)(x_1-x_3)}, \dots$$

$$f'(x) \approx L'(x) = L_1'(x)y_1 + L_2'(x)y_2 + L_3'(x)y_3$$

$$L_1'(x) = \frac{2x - x_2 - x_3}{(x_1 - x_2)(x_1 - x_3)}, \dots$$

- Example: Three-point forward difference formula

$$x_1 = x_0, x_2 = x_0 + h, x_3 = x_0 + 2h$$

$$L'(x_1) = \frac{-3f(x_1) + 4f(x_1 + h) - f(x_1 + 2h)}{2h}$$

$$f'(x_0) \approx \frac{-f(x_0 + 2h) + 4f(x_0 + h) - 3f(x_0)}{2h}$$

- Other approximations:  
Spline, etc.



# Richardson's Derivative Approximation

- Richardson extrapolation:

$$\varphi(h) = \frac{1}{2h} [f(x_0+h) - f(x_0-h)]$$

$$\varphi(h) = f(x_0) + a_2 h^2 + a_4 h^4 + a_6 h^6 + \dots$$

- Neville's tableaux

n\m	0	1	2
0	D(0,0)		
1	D(1,0)	D(1,1)	
2	D(2,0)	D(2,1)	D(2,2)

$$f(x_0) = \frac{4\varphi\left(\frac{h}{2}\right) - \varphi(h)}{3} + O(h^4)$$

$$D(n, 0) = \varphi\left(\frac{h}{2^n}\right)$$

$$D(n, m) = \frac{4^m D(n, m-1) - D(n-1, m-1)}{4^m - 1}$$

- Ridders' implementation

Compute  $D(n,n)$  until

$$|D(n+1, n+1) - D(n, n)| \geq |D(n, n) - D(n-1, n-1)|, \text{ or}$$

$$|D(n+1, n+1) - D(n, n)| \leq \text{tolerance}, \text{ or}$$

$n+1 > \text{max number .of .steps}$

# Complex-step Derivative Approximation

## ■ Complex-step derivative

$$f'(x_0) = \frac{\text{Im}(f(x_0 + h \cdot i))}{h} + O(h^2)$$

## ■ Remark

**Almost no loss of accuracy  
(no roundoff error)**

## ■ Proof

- Complex Taylor series

## ■ Assumptions

- $f$  can be analytically continued into a neighborhood of  $x_0$
- $f(x_0)$  is real
- $0 < h \ll 1$  real

## ■ Settings

- $\varepsilon_{f'} = \varepsilon_f$
- $h$  can be arbitrarily small  
e.g. take  $h = \varepsilon_m$

## ■ Recent successes (since 2000):

- Aerodynamics
- Weather forecast
- Molecular modeling

# Other Approaches

## ■ Automated Differentiation

Automated or Automatic Differentiation (AD) is a method for automatically augmenting or wrapping numerical procedures with statements for computing their derivatives. Especially useful in combination with complex-step derivative approximation.

## ■ Symbolic Differentiation

Symbolic differentiation is provided by most Computer Algebra Systems (CAS). Such a program finds the derivative of a given formula w.r.t. a specified variable, producing a new formula as output. Together with high-precision arithmetic, this approach delivers exact derivatives.

# Software and Literature

## ■ Software

### ■ Matlab

**Richardson extrapolation  
automated differentiation  
(from Mathtools.net)**

### ■ Octave

### ■ Scilab

### ■ *R*

**Richardson extrapolation  
and in optimization routines**

### ■ Euler, w/ Maxima

**Arbitrary-precision arithmetic  
symbolic algebra**

### ■ Netlib

Fortran/C implementations

## ■ Literature

Conte & de Boor, Elementary Numerical Analysis, McGraw-Hill, 1980.

Fausett, Applied Numerical Analysis Using Matlab. 2nd Edition, Prentice Hall, 2007.

Martins, Sturdza, and Alonso, The Complex-step Derivative Approximation. ACM ToMS, Vol. 29, No. 3, 2003, pp. 245-262.

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