

# Numerical Integration and Differentiation

## Computational Economics

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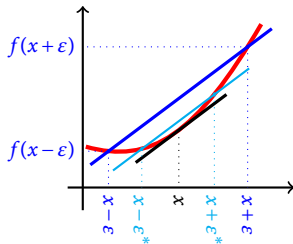
# Finite Differences for Numerical Differentiation

## basic concepts

- difference quotient:  $\frac{\Delta y}{\Delta x} = \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x}$
- derivative:  $\frac{dy}{dx} \equiv f'(x) \equiv \lim_{\Delta x \rightarrow 0} \frac{\Delta y}{\Delta x}$

## numerical approximation of $f'$ with central difference

$$\begin{aligned} f'(x) &\approx \frac{f(x + \varepsilon) - f(x - \varepsilon)}{(x + \varepsilon) - (x - \varepsilon)} \\ &= \frac{f(x + \varepsilon) - f(x - \varepsilon)}{2\varepsilon} \quad \text{where } \varepsilon \rightarrow 0 \end{aligned}$$



## multi-dimensional functions

- gradient: list of all first order derivatives

$$\begin{aligned} \nabla f &= \left[ \frac{\partial f}{\partial x_1} \quad \dots \quad \frac{\partial f}{\partial x_n} \right] \\ &\approx \left[ \frac{f(x_1 + \varepsilon, x_2, \dots, x_n) - f(x_1 - \varepsilon, x_2, \dots, x_n)}{2\varepsilon} \quad \dots \quad \frac{f(x_1, x_2, \dots, x_n + \varepsilon) - f(x_1, x_2, \dots, x_n - \varepsilon)}{2\varepsilon} \right] \end{aligned}$$

# Finite Differences for Numerical Differentiation

## second derivative for one-dimensional functions

- since second derivative = derivative of first derivative, compute by iteratively applying definitions:

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

where  $\epsilon = 2\epsilon$

- likewise for higher derivatives

# Finite Differences for Numerical Differentiation

## second derivative for multi-dimensional functions

- Hessian matrix multi-dimensional functions: matrix of all second order derivatives

$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \vdots & \frac{\partial^2 f}{\partial x_i \partial x_j} & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

- usually, Hessian is symmetric, i.e.,  $\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}$

# Finite Differences for Numerical Differentiation

## numerical approximation of the Hessian

- vary values in dimensions  $i$  and  $j$  (leaving all other values unchanged) and evaluate points

$$\begin{array}{ccc} & x_j - \varepsilon & x_j & x_j + \varepsilon \\ \begin{array}{c} x_i - \varepsilon \\ x_i \\ x_i + \varepsilon \end{array} & \begin{pmatrix} f^{--} & f^{-\circ} & f^{-+} \\ f^{\circ-} & f^{\circ\circ} & f^{\circ+} \\ f^{+-} & f^{+\circ} & f^{++} \end{pmatrix} \end{array}$$

- diagonal elements of the Hessian matrix

$$\frac{\partial^2 f}{\partial x_i^2} \approx \begin{cases} \frac{f^{+\circ} + f^{-\circ} - 2f^{\circ\circ}}{\varepsilon^2} & \text{alternative 1} \\ \frac{1}{2} \left( \frac{f^{+-} + f^{-+} - 2f^{\circ\circ}}{\varepsilon^2} + \frac{f^{++} + f^{--} - 2f^{\circ\circ}}{\varepsilon^2} \right) & \text{alternative 2} \end{cases}$$

- off-diagonal elements of the Hessian matrix (cross-derivatives)

$$\frac{\partial^2 f}{\partial x_i \partial x_j} \approx \begin{cases} \frac{1}{2} \left( \frac{f^{+\circ} + f^{-\circ} - f^{\circ+} - f^{\circ-}}{\varepsilon^2} + \frac{f^{+\circ} + f^{\circ-} - f^{+-} - f^{\circ\circ}}{\varepsilon^2} \right) & \text{alternative 1} \\ \frac{1}{2} \left( \frac{f^{\circ\circ} + f^{++} - f^{\circ+} - f^{+\circ}}{\varepsilon^2} + \frac{f^{\circ\circ} + f^{--} - f^{\circ-} - f^{-\circ}}{\varepsilon^2} \right) & \text{alternative 2} \\ \frac{1}{4} \left( \frac{f^{++} + f^{--} - f^{+-} - f^{-+}}{\varepsilon^2} \right) & \text{average alt.s 1 and 2} \end{cases}$$

# Application Example

## local approximation with Taylor Series Approximation

- approximation for region around a given value  $x_0$
- approximate function value at  $x = x_0 + \delta$  with a polynomial

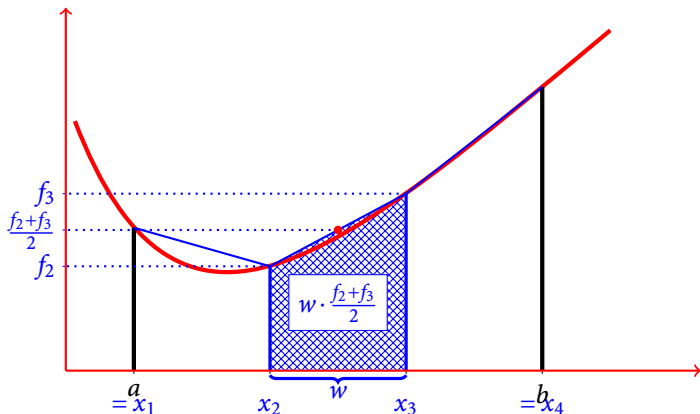
$$f(x) = f(x_0 + \delta) = f(x_0) + \delta f'(x_0) + \frac{\delta^2}{2} f''(x_0) + \frac{\delta^3}{6} f'''(x_0) + \dots \\ + \dots + \frac{\delta^n}{n!} f^{(n)}(x_0) + R_n$$

- remainder  $R_n < |x - x_0|^{n+1}$

# Numerical Integration with Trapezoid Rule

## basic idea

- assume function is approximately piecewise linear
- area underneath each line segment is a trapezoid
- integral is the sum of all trapezoids' areas



# Numerical Integration with Trapezoid Rule

how solve a problem like  $\int_a^b f(x)dx$

- split range into  $n$  segments of equal width  $w = (b - a)/n$
- compute borders of segments,  $x_i = a + (i - 1)w$  for  $i = 1..(n + 1)$
- approximation for area of one segment:  $\int_{x_i}^{x_{i+1}} f(x)dx \approx w \frac{f(x_i) + f(x_{i+1})}{2}$
- for the entire range (with  $f_i = f(x_i)$ )

$$\begin{aligned}\int_a^b f(x)dx &= \sum_{i=1}^n \int_{x_i}^{x_{i+1}} f(x)dx \\ &\approx \sum_{i=1}^n w \cdot \frac{f_i + f_{i+1}}{2} = w \cdot \left( \frac{f_1}{2} + f_2 + \dots + f_n + \frac{f_{n+1}}{2} \right)\end{aligned}$$

- more generally (when segments have different width):

$$\int_a^b f(x)dx \approx \sum_{i=1}^n (x_{i+1} - x_i) \cdot \frac{f_i + f_{i+1}}{2}$$

- extension: Simpson's rule (piecewise quadratic functions)



# Numerical Integration with Monte Carlo Simulation

some (very) simple examples

$$f(x) = \frac{1}{2}x \implies \int_0^2 f(x)dx =? \quad f(x) = \sin(|x|^{\sqrt[5]{|x|}}) \implies \int_{-1}^3 f(x)dx =?$$

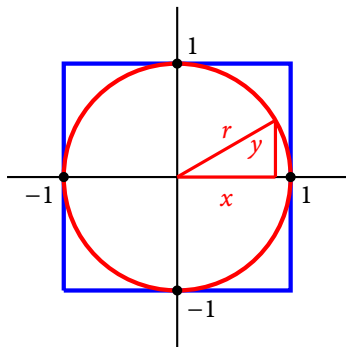
Estimating the Value of  $\pi$  via MCS

$$A_{\square} = (2r)^2$$

$$A_{\circ} = r^2 \pi$$

$$\frac{A_{\circ}}{A_{\square}} = \frac{r^2 \pi}{4r^2} = \frac{\pi}{4}$$

$$\implies \pi = 4 \frac{A_{\circ}}{A_{\square}}$$



# additional literature



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