Section 4.1 Numerical Differentiation

Motivation.

- Consider to solve Black-Scholes equation $\frac{\partial f}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 f}{\partial S^2} + rS \frac{\partial f}{\partial S} rf = 0$. Here f is the price of a derivative security, t is time, S is the varying price of the underlying asset, r is the risk-free interest rate, and σ is the market volatility.
- The heat equation of a plate: $\frac{\partial u}{\partial t} = k \left(\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$. Here k is the heat-diffusivity coefficient.

Goal: Compute accurate approximation to the derivative(s) of a function.

The derivative of
$$f$$
 at x_0 is: $f'(x_0) = \lim_{h \to 0} \frac{f(x_0+h)-f(x_0)}{h}$.

Obviously, when h is small, $\frac{f(x_0+h)-f(x_0)}{h}$ is a "good" approximation to $f'(x_0)$.

What is the error of approximation?

Big idea:

Build an interpolating polynomial to approximate f(x), then use the derivative of the interpolating polynomial as the approximation of the $f'(x_0)$.

Example: Consider to approximate $f'(x_0)$ using two points x_0 and $x_0 + h$.

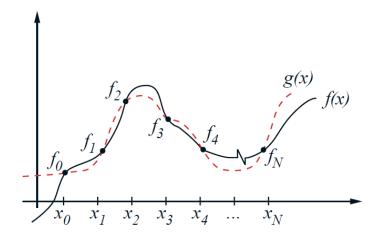
Example 4.4.1 Use forward difference formula with h = 0.1 to approximate the derivative of $f(x) = \ln(x)$ at $x_0 = 1.8$. Determine the bound of the approximation error.

Forward-difference:
$$f'(x_0) \approx \frac{f(x_0+h)-f(x_0)}{h}$$
 when $h > 0$.

Backward-difference:
$$f'(x_0) \approx \frac{f(x_0+h)-f(x_0)}{h}$$
 when $h < 0$.

1st derivative approximation (obtained by Lagrange interpolation)

The interpolation points are given as:



$$(x_0, f(x_0))$$

$$(x_1, f(x_1))$$

$$(x_2, f(x_2))$$

. . .

$$(x_N, f(x_N))$$

By Lagrange Interpolation Theorem (**Thm 3.3**):

$$f(x) = \sum_{k=0}^{n} f(x_k) L_{N,k}(x) + \frac{(x - x_0) \cdots (x - x_N)}{(N+1)!} f^{(N+1)}(\xi(x))$$
 (1)

Take 1st derivative for Eq. (1):

$$f'(x) = \sum_{k=0}^{n} f(x_k) L'_{N,k}(x) + \frac{(x - x_0) \cdots (x - x_N)}{(N+1)!} \left(\frac{d \left(f^{(N+1)} (\xi(x)) \right)}{dx} \right) + \frac{1}{(N+1)!} \left(\frac{d \left((x - x_0) \cdots (x - x_N) \right)}{dx} \right) f^{(N+1)} (\xi(x))$$

Set $x = x_j$, with x_j being x-coordinate of one of interpolation nodes. j = 0, ..., N.

$$f'(x_j) = \sum_{k=0}^{n} f(x_k) L'_{N,k}(x_j) + \frac{f^{(N+1)}(\xi(x_j))}{(N+1)!} \prod_{\substack{k=0; \\ k \neq j}}^{N} (x_j - x_k)$$

---- (N+1)-point formula with error to approximate $f'(x_i)$ (4.2)

The error of (N+1)-point formula is $\frac{f^{(N+1)}(\xi(x_j))}{(N+1)!}\prod_{\substack{k=0;\\k\neq j}}^{N}(x_j-x_k).$

Remark: $f'(x_j) \approx \sum_{k=0}^n f(x_k) L'_{N,k}(x_j)$

Example. Derive the three-point formula with error to approximate $f'(x_j)$. Let interpolation nodes be $(x_0, f(x_0)), (x_1, f(x_1))$ and $(x_2, f(x_2))$.

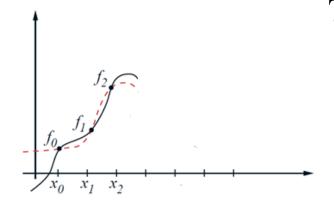
$$f'(x_j) = f(x_0) \left[\frac{2x_j - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)} \right] + f(x_1) \left[\frac{2x_j - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)} \right]$$

$$+ f(x_2) \left[\frac{2x_j - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)} \right] + \frac{f^{(3)}(\xi(x_j))}{6} \prod_{\substack{k=0; \\ k \neq j}}^{2} (x_j - x_k)$$

Mostly used three-point formula (see Figure 1)

Let x_0 , x_1 , and x_2 be equally spaced and the grid spacing be h.

Thus $x_1 = x_0 + h$; and $x_2 = x_0 + 2h$.



$$1.f'(x_0) = \frac{1}{2h} \left[-3f(x_0) + 4f(x_1) - f(x_2) \right] + \frac{h^2}{3} f^{(3)} \left(\xi(x_0) \right)$$
 (three-point endpoint formula with error) (4.4)

2.
$$f'(x_1) = \frac{1}{2h} \left[-f(x_0) + f(x_2) \right] + \frac{h^2}{6} f^{(3)} (\xi(x_1))$$
 (three-point midpoint formula with error) (4.5)

$$3.f'(x_2) = \frac{1}{2h} [f(x_0) - 4f(x_1) + 3f(x_2)] + \frac{h^2}{3} f^{(3)}(\xi(x_2))$$
(three-point endpoint formula with error) (4.4.1)

Remark: Eq. (4.4) in textbook is:

$$f'(x_0) = \frac{1}{2h} \left[-3f(x_0) + 4f(x_0 + h) - f(x_0 + 2h) \right] + \frac{h^2}{3} f^{(3)} \left(\xi(x_0) \right)$$

h can be both positive and negative

Mostly used five-point formula

1. Five-point midpoint formula

$$f'(x_0) = \frac{1}{12h} [f(x_0 - 2h) - 8f(x_0 - h) + 8f(x_0 + h) - f(x_0 + 2h)] + \frac{h^4}{30} f^{(5)}(\xi)$$
(4.6)

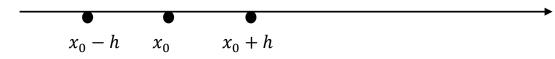
2. Five-point endpoint formula

$$f'(x_0) = \frac{1}{12h} \left[-25f(x_0) + 48f(x_0 + h) - 36f(x_0 + 2h) + 16f(x_0 + 3h) - 3f(x_0 + 4h) \right] + \frac{h^4}{5} f^{(5)}(\xi)$$
(4.7)

Example 4.1.2 Values for $f(x) = xe^x$ are given in the following table. Use all applicable 3-point and 5-point formulas to approximate f'(2.0).

χ	1.8	1.9	2.0	2.1	2.2
f(x)	10.889365	12.703199	14.778112	17.148957	19.855030

2nd derivative approximation (obtained by Taylor polynomial)



Approximate $f(x_0 + h)$ by expansion about x_0 :

$$f(x_0 + h) = f(x_0) + f'(x_0)h + \frac{1}{2}f''(x_0)h^2 + \frac{1}{6}f'''(x_0)h^3 + \frac{1}{24}f^{(4)}(\xi_1)h^4$$
(3)

Approximate $f(x_0 - h)$ by expansion about x_0 :

$$f(x_0 - h) = f(x_0) - f'(x_0)h + \frac{1}{2}f''(x_0)h^2 - \frac{1}{6}f'''(x_0)h^3 + \frac{1}{24}f^{(4)}(\xi_2)h^4$$
(4)

Add Eqns. (3) and (4):
$$f(x_0 - h) + f(x_0 + h) = 2f(x_0) + f''(x_0)h^2 + \left[\frac{1}{24}f^{(4)}(\xi_1)h^4 + \frac{1}{24}f^{(4)}(\xi_2)h^4\right]$$

Thus

Second derivative midpoint formula

$$f''(x_0) = \frac{1}{h^2} [f(x_0 - h) - 2f(x_0) + f(x_0 + h)] - \frac{h^2}{12} f^{(4)}(\xi)$$

Example 3. Values for $f(x) = xe^x$ are given in the following table. Use second derivative approximation formula to approximate f''(2.0).

х	1.8	1.9	2.0	2.1	2.2
f(x)	10.889365	12.703199	14.778112	17.148957	19.855030

Solution:
$$f''(2.0) \approx \frac{1}{(0.1)^2} [f(1.9) - 2f(2.0) + f(2.1)] =$$

Or

$$f''(2.0) \approx \frac{1}{(0.2)^2} [f(1.8) - 2f(2.0) + f(2.2)] =$$

Round-Off Error Instability

Let the round-off errors associated with $f(x_0 + h)$ and $f(x_0 - h)$ be $e(x_0 + h)$ and $e(x_0 - h)$, respectively.

Then
$$f(x_0 + h) = \tilde{f}(x_0 + h) + e(x_0 + h);$$

$$f(x_0 - h) = \tilde{f}(x_0 - h) + e(x_0 - h).$$

Here $\tilde{f}(x_0 + h)$ and $\tilde{f}(x_0 - h)$ are actual values used by computer.

The total error of approximation using three-point midpoint formula:

$$f'(x_0) - \frac{\tilde{f}(x_0 + h) - \tilde{f}(x_0 - h)}{2h} = \frac{e(x_0 + h) - e(x_0 - h)}{2h} - \frac{h^2}{6}f^{(3)}(\xi_1)$$

Assume round-off errors are bounded by $\varepsilon \ge 0$, $|f^{(3)}(\xi_1)| \le M$.

Then:

$$\left|f'(x_0) - \frac{\tilde{f}(x_0 + h) - \tilde{f}(x_0 - h)}{2h}\right| \le \frac{\varepsilon}{h} + \frac{h^2}{6}M.$$

Remark:

- 1. As h reduces, $\frac{\varepsilon}{h}$ grows;
- 2.In practice, it's rare to let *h* be too small;
- 3. Let the total error be $\frac{\varepsilon}{h} + \frac{h^2}{6}M$, a minimum of the total error occurs at

$$h = \left(\frac{3\varepsilon}{M}\right)^{\frac{1}{3}}.$$