

2/1/2010

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#9

$$x = 0.0001852$$

$$y = 0.0003899$$

Wrong way:

$$x = 0.0002$$

$$y = 0.0004$$

$$\Rightarrow x + y = 0.0006 = 0.6 \times 10^{-3} = z_1$$

Correct way: use floating point representation

$$x = 0.1852 \times 10^{-3}$$

$$y = 0.3899 \times 10^{-3}$$

$$x + y = (0.1852 + 0.3899) \times 10^{-3} =$$

$$= \underbrace{0.5651}_{\text{}} \times 10^{-3} = z_2$$

$$z_1 \neq z_2!$$

Newton's method (Cont'd)

last time we showed that provided $f'(d) \neq 0$

$$d - x_{n+1} = -\frac{1}{2} \frac{f''(\xi_n)}{f'(x_n)} (d - x_n)^2 \approx$$

d : exact

$$\approx -\frac{1}{2} \frac{f''(d)}{f'(d)} (d - x_n)^2 = M (d - x_n)^2$$

where

$$M = -\frac{1}{2} \frac{f''(d)}{f'(d)}$$

$$\Rightarrow d - x_{n+1} = M \cdot (d - x_n)^2 \quad | \quad M$$

$$M(d - x_{n+1}) = (M(d - x_n))^2 = [M(d - x_{n-1})]^2 = \dots$$

$$\dots = [M(d - x_0)]^{2^{n+1}}$$

d : root, x_0 : initial guess, M : const

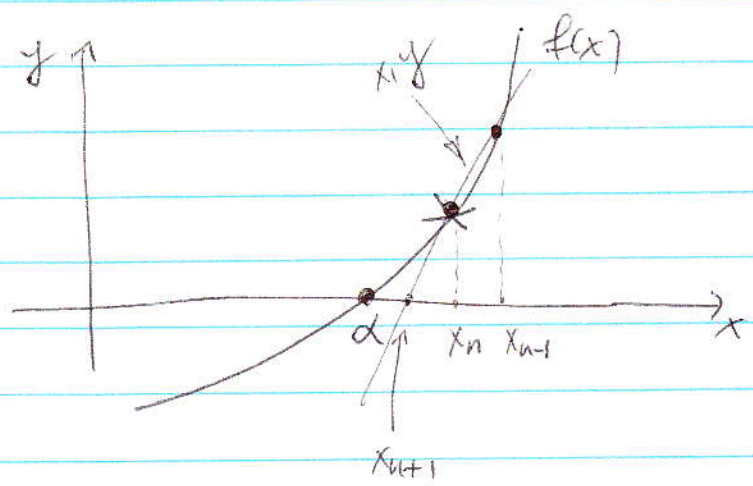
For convergence we need $|M(d - x_0)| < 1$

If $|M|$ is large, then x_0 may need to be chosen very closely to root d for convergence.

Secant method

$$x_{n+1} = x_n - \frac{f(x_n)}{\frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}} \quad \text{given } x_0, x_1$$

Geometrically



Slope of secant line: $\frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$

Equation of secant line:

$$y = ax + b$$

$$y = kx + b$$

$$y - y_0 = k(x - x_0)$$

$$\frac{y - y_0}{x - x_0} = k$$

$$\frac{y - f(x_n)}{x - x_n} = \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$$

Set $y=0$ and $x=x_{n+1} \Rightarrow$ solve for x_{n+1} : as above

Note

1. It can be shown that $|x - x_n| \leq C \cdot |x - x_{n+1}|^p$

where $p = \frac{1 + \sqrt{5}}{2} \sim 1.6$ golden ratio

superlinear convergence

2. Secant method converges more slowly than Newton's method but faster than fixed-point iterations (in general).

Summary

1. Bisection

- linear convergence
- guaranteed to converge
- do not need $f'(x)$

2. Secant

- superlinear convergence
- convergence depends on initial guesses x_0 and x_1
- do not need $f'(x)$

3. Newton's method

- converges quadratically (if root α is simple)
- convergence depends on x_0
- need to know $f'(x)$

Linear systems of equations

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n$$

n equations, n unknowns x_1, x_2, \dots, x_n

$$\sum_{j=1}^n a_{ij}x_j = b_i \quad i=1, \dots, n$$

$$\underbrace{\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}}_A \underbrace{\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}}_x = \underbrace{\begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}}_b$$

$$\Rightarrow Ax = b, \quad A = (a_{ij})$$

Thm

Let A be an $n \times n$ matrix. The following conditions on A are equivalent:

1. The equation $Ax = b$ has a unique solution for any b .
2. $\det A \neq 0$
3. Matrix A is invertible, i.e. there is matrix that we denote A^{-1} such that $AA^{-1} = A^{-1}A = \underline{I}$
 I : identity matrix
 A^{-1} : inverse matrix
4. Columns of A are linearly independent
5. Rows of A are linearly independent
6. Eigenvalues of A are all nonzero.
7. The equation $Ax = 0$ has a unique solution $x = 0$

Direct methods

$Ux = b$ where U is upper triangular matrix

$$u_{11}x_1 + u_{12}x_2 + \dots + u_{1n}x_n = b_1$$

$$u_{22}x_2 + \dots + u_{2n}x_n = b_2$$

$$u_{n-1,n-1}x_{n-1} + u_{n-1,n}x_n = b_{n-1}$$

$$u_{nn}x_n = b_n$$

Then

$$x_n = b_n / u_{nn}$$

$$x_{n-1} = (b_{n-1} - u_{n-1,n}x_n) / u_{n-1,n-1}$$

$$x_1 = (b_1 - (u_{12}x_2 + \dots + u_{1n}x_n)) / u_{11}$$

This is called back substitution.

Operation count

divisions := n

multiplications = $n(n-1)/2$

Pf

$$\# \text{ mult} = 1 + 2 + \dots + n-1 = \sum_{k=1}^{n-1} k = S$$

$$2S = \sum_{k=1}^{n-1} k + \sum_{k=1}^{n-1} (n-k) = \sum_{k=1}^{n-1} (k + (n-k)) = \sum_{k=1}^{n-1} n$$

$$= n(n-1)$$

or

The leading order term in operation count is $\frac{n^2}{2}$