

$\sin x = x - x^3/3! + x^5/5! - \dots$
 $\cos x = 1 - x^2/2! + x^4/4! - \dots$
 $e^x = 1 + x + x^2/2! + x^3/3! + \dots$
 $\ln x = (x-1) - (x-1)^2/2 + (x-1)^3/3 - \dots$
 $\ln(1+x) = x - (1/2)x^2 + (1/3)x^3 - \dots$

Weighted Mean value theorem:
 $\int f(x)g(x) = f(c) \int g(x)$

$g \in C[a,b]$ and $g'(x) \leq k < 1 \Rightarrow$
 converge to unique fix point from
 any point in $[a,b]$

$f \in C^2[a,b]$ and $f(p)=0$ and
 $f'(p) \neq 0 \Rightarrow$ converge for some
 initial approx around p

$g \in C^3[a,b]$ and $x=g(x)$ has sol'n p
 and $g'(p) \neq 1 \Rightarrow$ converge for
 initial approx in $[a,b]$ using
 Stevenson's method

Order of convergence

$$\lim_{n \rightarrow \infty} \frac{|p_{n+1} - p|}{|p_n - p|^a} = \lambda$$

Aiken's method

$$\hat{p} = p_n - \frac{(\Delta p_n)^2}{\Delta^2 p_n}$$

Lagrange:

$$P(x) = \sum_{k=0}^n f(x_k) L_{n,k}(x),$$

$$L_{n,k} = \prod_{i=0, i \neq k}^n \frac{x - x_i}{x_k - x_i}$$

Neville's method

$$\frac{(x - x_j)P_{no j}(x) - (x - x_i)P_{no i}(x)}{(x_i - x_j)}$$

x_0	$P_0=Q_{0,0}$		
x_1	$P_1=Q_{1,0}$	$P_{0,1}=Q_{1,1}$	
x_2	$P_2=Q_{2,1}$	$P_{1,2}=Q_{2,1}$	$P_{0,1,2}=Q_{2,2}$

Cubic Spline

- a) $S_j(x)$ is cubic on subinterval $[x_j, x_{j+1}]$ for each $j=0,1,\dots,n-1$
- b) $S(x_j)=f(x_j)$
- c) $S_{j+1}(x_{j+1})=S_j(x_{j+1})$
- d) $S'_{j+1}(x_{j+1})=S'_j(x_{j+1})$
- e) $S''_{j+1}(x_{j+1})=S''_j(x_{j+1})$
- f) **Natural/Free:** $S''(x_0)=S''(x_n)=0$
- Clamped:** $S'(x_0)=F'(x_0)$ and $S'(x_n)=F'(x_n)$

Lagrange Error:

$$f(x) = P(x) + \frac{f^{(n+1)}(\xi(x))}{(n+1)!} (x-x_0)(x-x_1)\dots(x-x_n)$$

Proof: Consider $g(t) = f(t) - P(t) - [f(x) - P(x)] \prod_{i=0}^n \frac{t-x_i}{x-x_i}$,
 which has $n+2$ roots $\Rightarrow \exists \xi$ s.t. $g^{(n+1)}(\xi)=0$

Divided difference

k th divided difference:

$$f[x_i, \dots, x_{i+k}] = \frac{f[x_{i+1}, \dots, x_{i+k}] - f[x_i, \dots, x_{i+k-1}]}{x_{i+k} - x_i}$$

$$P_n(x) = f[x_0] + \sum_{k=1}^n f[x_0, x_i, \dots, x_k](x-x_0)\dots(x-x_{k-1})$$

$$= \sum_{k=0}^n s(s-1)\dots(s-k+1)h^k f[x_0, x_i, \dots, x_k]$$

x	$f(x)$	First divided differences	Second divided differences	Third divided differences
x_0	$f[x_0]$	$f[x_0, x_1] = \frac{f[x_1] - f[x_0]}{x_1 - x_0}$		
x_1	$f[x_1]$	$f[x_1, x_2] = \frac{f[x_2] - f[x_1]}{x_2 - x_1}$	$f[x_0, x_1, x_2] = \frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0}$	
x_2	$f[x_2]$	$f[x_2, x_3] = \frac{f[x_3] - f[x_2]}{x_3 - x_2}$	$f[x_1, x_2, x_3] = \frac{f[x_2, x_3] - f[x_1, x_2]}{x_3 - x_1}$	$f[x_0, x_1, x_2, x_3] = \frac{f[x_1, x_2, x_3] - f[x_0, x_1, x_2]}{x_3 - x_0}$
x_3	$f[x_3]$	$f[x_3, x_4] = \frac{f[x_4] - f[x_3]}{x_4 - x_3}$	$f[x_2, x_3, x_4] = \frac{f[x_3, x_4] - f[x_2, x_3]}{x_4 - x_2}$	$f[x_1, x_2, x_3, x_4] = \frac{f[x_2, x_3, x_4] - f[x_1, x_2, x_3]}{x_4 - x_1}$
x_4	$f[x_4]$	$f[x_4, x_5] = \frac{f[x_5] - f[x_4]}{x_5 - x_4}$	$f[x_3, x_4, x_5] = \frac{f[x_4, x_5] - f[x_3, x_4]}{x_5 - x_3}$	$f[x_2, x_3, x_4, x_5] = \frac{f[x_3, x_4, x_5] - f[x_2, x_3, x_4]}{x_5 - x_2}$
x_5	$f[x_5]$			

Spectral Radius and Iterative Methods

(1) $\|A\|_2 = [p(A^t A)]^{1/2}$ (2) $\rho(A) \leq \|A\|$ for any $\|\cdot\|$

Following statements are equivalent:

- (1) A is convergent matrix
- (2) $\|A^n\| = 0$ as $n \rightarrow \infty$ for some $\|\cdot\|$
- (3) $\rho(A) < 1$
- (4) $\|A^n\| = 0$ as $n \rightarrow \infty$ for all $\|\cdot\|$
- (5) $A^n x = 0$ as $n \rightarrow \infty$ for all x

Th'm: If $\rho(T) < 1$, then $(1-T)^{-1}$ exists and $(1-T)^{-1} = 1 + T + T^2 + \dots$

Proof: (1) $\det(1-T) = 0 \Rightarrow 1$ is an eigenvalue
 (2) $(1-T)S_m = 1 + T^2 + \dots + T^{m-1}$. Let $m \rightarrow \infty$.

Th'm: The sequence $\{x^{(k)}\}_{k=0}^{\infty}$ defined by $x^{(k)} = Tx^{(k-1)} + c$

converges to unique sol'n $x = Tx + c$ iff $\rho(T) < 1$

Proof: (\Leftarrow) $x^{(k)} = Tx^{(k-1)} + c = T(Tx^{(k-2)} + c) + c = \dots = T^k x^{(0)} + (T^{k-1} + \dots + T + I)c$. $k \rightarrow \infty \Rightarrow 0 + (1-T)^{-1}c$
 (\Rightarrow) For any z , construct $x^{(0)} = x - z$. Assume $\{x^{(k)}\}$ converges to x . Note $x^{(k)} = (Tx + c) - (Tx^{(k-1)} + c) = T(x - x^{(k-1)})$.
 $\therefore x - x^{(k)} = T(x - x^{(k-1)}) = T^2(x - x^{(k-2)}) = \dots = T^k(x - x^{(0)}) = T^k z$. As $k \rightarrow \infty$, LHS approaches 0. $T^k z = 0$ for any $z \Rightarrow \rho(T) < 1$

Jacobi Iterative method: Write $A = D - L - U$

$$T_j = D^{-1}(L + U) \quad c_j = D^{-1}b \quad x^{(k)} = T_j x^{(k-1)} + c_j$$

Gauss-Seidel: Write $A = D - L - U$

$$T_g = (D - L)^{-1}U \quad c_g = (D - L)^{-1}b \quad x^{(k)} = T_g x^{(k-1)} + c_g$$

$$x_i^{(k)} = \frac{-\sum_{j=1}^{i-1} (a_{ij} x_j^{(k)}) - \sum_{j=i+1}^n (a_{ij} x_j^{(k-1)}) + b_i}{a_{ii}}$$

Error Bound: $\|x - x^{(k)}\| \leq \|T\|^k \|x^{(0)} - x\| = (\|T\|^k / (1 - \|T\|)) \|x^{(1)} - x^{(0)}\|$

Definition of Residual Vector: $r = b - A\tilde{x}$

Strictly diag. dom.: $|a_{ii}| > \sum_{j=1, j \neq i}^n |a_{ij}|$

- (1) Strictly diag. matrix is nonsingular.
- (2) Gauss. elim. can be performed without row/column interchanges
- (3) Computation will be stable with respect to round off error
- (4) **Has LU factorization**

Positive Definite:

A is positive definite if it is

- (1) symmetric and
- (2) $x^t A x > 0$ for $x \neq 0$

A is positive definite matrix **then**

- (1) A is nonsingular
 - (2) $a_{ii} > 0$, for each $i=1,2,\dots,n$
 - (3) $\max_{i \leq k, j \leq n} |a_{kj}| \leq \max_{1 \leq i \leq n} |a_{ii}|$
 - (4) $(a_{ij})^2 < a_{ii} a_{jj}$ for each $i \neq j$
 - (5) All eigenvalues are positive
 - (6) GE solves without row-swap (All pivot element > 0). Can write $A = LU$
- Matrix A can be factored into LDL^t or LL^t iff A is positive definite**

Leading Principle submatrix

$\begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$ has 2, $\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$ and itself

Th'm: A symmetric matrix A is positive definite iff each of its leading principle submatrices has a positive determinant.

Permutation Matrix:

$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} A \Rightarrow \begin{cases} \text{1st row of } A \\ \text{3rd row of } A \\ \text{2nd row of } A \end{cases}$

Properties of Vector Norm

- (i) $\|x\| \geq 0$
- (ii) $\|x\| = 0 \Leftrightarrow x = 0$
- (iii) $\|ax\| = |a| \|x\|$
- (iv) $\|x+y\| \leq \|x\| + \|y\|$

$\|x\|_2 = \text{Euclidean}$ $\|x\|_{\infty} = \max |x_i|$

Th'm: The sequence of vectors $\{x^{(k)}\}$ converges to x with respect to $\|\cdot\|_{\infty}$ iff

$\lim_{k \rightarrow \infty} x_i^{(k)} = x_i$

Th'm: $\|x\|_{\infty} \leq \|x\|_2 \leq \sqrt{n} \|x\|_{\infty}$

Matrix Norm $\|A\| = \max_{\|x\|=1} \|Ax\|$

- (i) $\|A\| \geq 0$
- (ii) $\|A\| = 0 \Leftrightarrow A = 0$
- (iii) $\|aA\| = |a| \|A\|$
- (iv) $\|AB\| \leq \|A\| \|B\|$
- (v) $\|A+B\| \leq \|A\| + \|B\|$

$\|x\|_2 = \text{Euclidean}$ $\|x\|_{\infty} = \max |x_i|$

$$\|A\|_{\infty} = \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}|$$

Numerical Differentiation (unstable)

Forward diff.: $\frac{f(x_0+h)-f(x_0)}{h} - \frac{h}{2} f''(\xi)$ (Backward=h<0)

n+1-point formula: $x_0 \leq \xi_1 \leq x_0+2h$; $x_0-h \leq \xi_2 \leq x_0+h$

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

3-point formula: (End point)

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

3-point formula: (Mid point):

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

Mid-point with round-off, error term: $\frac{\epsilon}{h} + \frac{h^2}{6} M$,

ϵ : round off bound for each step; $M: 3^{rd}$ derivative bound

Derivation: Differentiating LaGrange with error term

Higher order derivatives: $x_0 \leq \xi_3 \leq x_0+2h$;

$$f''(x_0) = \frac{1}{h^2} (f(x_0-h) - 2f(x_0) + f(x_0+h)) + \frac{h^2}{12} f^{(4)}(\xi_3)$$

Derivation: Consider Taylors of $f(x_0+h)$ and $f(x_0-h)=f(x_0)-f'(x_0)h+(1/2)f''(x_0)h^2+\dots$. Find linear combinations of those two such that the coefficient for $f'(x_0)$ is 0, Rearrange for $f''(x_0)$.

Numerical Integration (stable):

Trapezoid Rule: $\frac{h}{2} [f(x_0) + f(x_1)] - \frac{h^3}{12} f''(\xi)$

Simpson's Rule: $\frac{h}{3} [f(x_0) + 4f(x_1) + f(x_2)] - \frac{h^5}{90} f^{(4)}(\xi)$

Def'n: Degree of accuracy / precision: largest n s.t. the approximation is exact for x^k , for $k \leq n$

Composite Simpson's: $\frac{h^5}{90} \sum_{j=1}^{n/2} f^{(4)}(\xi_j) = \frac{b-a}{180} h^4 f^{(4)}(\xi)$.

Total round off error: $(b-a)\epsilon$, ϵ is the bound for round-off.

Composite Trapezoid: $\frac{b-a}{12} h^2 f''(\xi)$

Adaptive Quadrature:

Theory

If $h=(a+b)/2$, then $S(a,b)$ has error $(h^5/90)f^{(4)}(e_1)$, and $(S(a,(a+b)/2)+S((a+b)/2,b))$ has error $(1/16)(h^5/90)f^{(4)}(e_2)$

$$\left| \int_a^b f(x)dx - S(a, \frac{a+b}{2}) - S(\frac{a+b}{2}, b) \right| \approx \frac{1}{15} \left| S(a,b) - S(a, \frac{a+b}{2}) - S(\frac{a+b}{2}, b) \right|$$

Procedure

Calculate $S(a,b)$, $S(a,(a+b)/2)$, $S((a+b)/2,b)$.

If $\frac{1}{15} \left| S(a,b) - S(a, \frac{a+b}{2}) - S(\frac{a+b}{2}, b) \right| > \text{tolerance}$,

repeat procedure on the subinterval with tolerance=tolerance/2

Initial Value Problem:

$f(t,y)$ satisfies **Lipschitz condition** in y on set $D \in \mathbb{R}^2$ if a constant $L > 0$ exists with $|f(t,y_1) - f(t,y_2)| \leq L|y_1 - y_2|$

With $\frac{dy}{dt} = f(t,y)$, $0 \leq t \leq 2$, $y(a) = \alpha$, If f is continuous and satisfies

Lipschitz in variable y , then the IVP is **well-posed** (unique sol'n exists and can be perturbed.).

Euler's Method: Difference Equation: $w_0=a$; $w_{i+1}=w_i+hf(t_i,w_i)$

Error: $|y(t_i) - w_i| \leq \frac{hM}{2L} [e^{L(t_i-a)} - 1]$

Difference Method: $w_0=a$; $w_{i+1}=w_i + \phi hf(t_i,w_i)$

Local Trunc. Err.:

$$T_{i+1}(h) = \frac{y_{i+1} - (y_i + h\phi(t_i, y_i))}{h} = \frac{y_{i+1} - y_i}{h} - \phi(t_i, y_i)$$

Euler local trunc. err: $T_{i+1}(h) = (h/2)y''(e) \Rightarrow O(h)$ for bounded y''

Higher order Taylor Methods:

$w_0=a$; $w_{i+1}=w_i+hT^{(n)}(t_i,w_i)$,

where $T^{(n)}(t_i, w_i) = f(t_i, w_i) + \frac{h}{2} f'(t_i, w_i) + \dots + \frac{h^{n-1}}{n!} f^{(n-1)}(t_i, w_i)$

Finding $y''(t)$: $f' = y''(t) = \frac{\partial f}{\partial t}(t, y(t)) + \frac{\partial f}{\partial y}(t, y(t)) \cdot f(t, y(t))$

Taylor Expansion of Two variables:

$$f(t + \alpha, y + \beta) = f(t, y) + \alpha \frac{\partial f}{\partial t}(t, y) + \beta \frac{\partial f}{\partial y}(t, y) + \frac{\alpha^2}{2} \frac{\partial^2 f}{\partial t^2}(\lambda, \gamma) + \alpha\beta \frac{\partial^2 f}{\partial t \partial y}(\lambda, \gamma) + \frac{\beta^2}{2} \frac{\partial^2 f}{\partial y^2}(\lambda, \gamma)$$

Midpoint Method $O(h^2)$

$w_0=a$; $w_{i+1}=w_i+h \cdot f(t_i+h/2, w_i + (h/2) \cdot f(t_i, w_i))$

Derivation: Matching coefficients of

$$T^{(2)}(t_{i+1}, w_{i+1}) = f(t_i, w_i) + \frac{h}{2} \frac{\partial f}{\partial t}(t_i, w_i) + \frac{h}{2} \frac{\partial f}{\partial y}(t_i, w_i) \cdot f(t_i, w_i) \text{ with}$$

$$a_1 f(t + \alpha_1, y + \beta_1) \Rightarrow a_1=1; a_1 \alpha_1 = h/2; a_1 \beta_1 = (h/2)f(t, y)$$

Modified Euler Method $O(h^2)$:

$w_0=a$; $w_{i+1} = w_i + \frac{h}{2} [f(t_i, w_i) + f(t_{i+1}, w_i + hf(t_i, w_i))]$

Derivation: By matching $T^{(2)}$ with $a_1 f(t, y) + a_2 f(t + \alpha, y + \beta f(t, y))$

Gauss. Elimination $O(n^3)$: $j = i+1 \dots n$; $m_{ji} = a_{ji} / a_{ii}$; $(E_j - m_{ji}E_i) \rightarrow E_j$;

$$*: \sum_1^{n-1} (n-i) + (n-i)(n-i+1) \quad +: \sum_1^{n-1} (n-i)(n-i+1)$$

Back substitution:

$$*: 1 + \sum_1^{n-1} (n-i) + 1 = \frac{n^2 + n}{2} \quad +: \sum_1^{n-1} (n-i-1) + 1 = \frac{n^2 - n}{2}$$

Partial pivot $O(n^2)$: choose element from column with largest absolute value and interchange

Scaled partial pivoting $O(n^2)$: Set s_i to be the element with largest absolute value in each row. When choosing a pivot, choose the element that has the largest absolute value when divided by the corresponding s_i in initial matrix.

Numerical Analysis Reference Sheet

(MACM 316, Simon Fraser University)

By Vincent Chu.

E-mail: chuvincen (at) gmail (dot) com

Reference based on Numerical Analysis, by R.L. Burden and J.D. Faires, Brookes-Cole, Seventh Edition.

Copyright 2005. All rights reserved.