

Number Systems

Base β

$$x = \pm(.a_1 a_2 a_3 \dots a_i \dots)_\beta \beta^e, \quad 1 \leq a_i < \beta$$

Chopping

$$\tilde{x} = \pm(.a_1 a_2 a_3 \dots a_i)_\beta \beta^e$$

Rounding

$$\tilde{x} = \begin{cases} \pm(.a_1 \dots a_i)_\beta \beta^e, & a_{i+1} < \frac{\beta}{2} \\ \pm[(.a_1 \dots a_i)_\beta \beta^e + (.0 \dots 1)_\beta \beta^e], & a_{i+1} \geq \frac{\beta}{2} \end{cases}$$

Error

Error: $e(\tilde{x}) = |x - \tilde{x}|$

Relative Error: $re(\tilde{x}) = \left| \frac{x - \tilde{x}}{x} \right|$

Linear Systems, $Ax = b$

THM: Given a matrix A , the following are equivalent

1. The Equation $Ax = b$ has a unique solution
2. A is invertible.
3. $\det(A) \neq 0$
4. $Ax = 0$ has a unique solution, $x = 0$
5. The columns of A are linearly independent
6. The eigenvalues, λ , of A are non-zero.

Gaussian Elimination: $A = LU$

Gaussian Elimination with pivoting: $PA = LU$

Norms

Properties of Vector Norms:

$$\|x\| \geq 0, \quad \|x\| = 0 \Rightarrow x = 0$$

$$\|\lambda x\| = |\lambda| \|x\|, \quad \lambda \text{ scalar}$$

$$\|x + y\| \leq \|x\| + \|y\|$$

Vector Norms:

$$l_\infty: \quad \|x\|_\infty = \max_{1 \leq i \leq n} |x_i|$$

$$l_1: \quad \|x\|_1 = \sum_{i=1}^n |x_i|$$

$$l_2: \quad \|x\|_2 = \left(\sum_{i=1}^n x_i^2 \right)^{\frac{1}{2}}$$

Matrix Norm:

$$\|A\| = \max_{\|u\| \neq 0} \{\|Au\| / \|u\| : u \in \mathbf{R}^n\}$$

Properties of Matrix Norms:

$$\|A\| \geq 0, \quad \|A\| = 0 \Leftrightarrow A = 0$$

$$\|\lambda A\| = |\lambda| \|A\|$$

$$\|A + B\| \leq \|A\| + \|B\|$$

$$\|Ax\| \leq \|A\| \|x\|$$

$$\|AB\| \leq \|A\| \|B\|$$

Examples of Matrix Norms:

$$l_\infty \text{ Matrix Norm:} \quad \|A\|_\infty = \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{i,j}|$$

$$l_1 \text{ Matrix Norm:} \quad \|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{i,j}|$$

$$l_2 \text{ Matrix Norm:} \quad \|A\|_2 = \sqrt{\rho(A^*A)}$$

Stability

Condition Number: $\kappa(A) = \|A^{-1}\| \|A\|$

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

THM:

$$\frac{1}{\kappa(A)} \frac{\|r\|}{\|b\|} \leq \frac{\|e\|}{\|x\|} \leq \kappa(A) \frac{\|r\|}{\|b\|}$$

Iterative Methods

$A = (L + D + U)$ where D is a diagonal matrix, L is lower triangular and U is upper triangular.

Jacobi: $Dx^{n+1} = -(L + U)x^n + b$

Gauss-Seidel: $Dx^{n+1} = -(Lx^{n+1} + Ux^n) + b$

SOR: $(D + \omega L)x^{n+1} = ((1 - \omega)D - \omega U)x^n + \omega b$

Root Finding Methods

Newton's Methods: $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$

Secant Methods: $x_{n+1} = x_n - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}$

Error Bound for Bisection Method:

$$|\alpha - x_n| \leq \left(\frac{1}{2}\right)^n |b_0 - a_0|$$

Polynomial Interpolation

Let f be defined on $[a, b]$; x_0, x_1, \dots, x_n : $n + 1$ distinct points in $[a, b]$. Let p_n be the interpolating polynomial of degree $\leq n$. Then

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

for some $\xi \in [a, b]$.

Chebyshev Points

$$x_k = \cos(2k+1)\pi/2n, \quad k = 0, 1, \dots, n-1$$

or

$$x_k = -\cos \pi kn, \quad k = 0, 1, \dots, n$$

Hermite Interpolation

Given f, x_0, x_1, \dots, x_n : $n+1$ distinct points, the Hermite interpolating polynomial $p(x)$ ($\deg p \leq 2n+1$) is

$$p(x) = \sum_{i=0}^n \left(f(x_i) h_i(x) + f'(x_i) \tilde{h}_i(x) \right)$$

where

$$h_i(x) = (1 - 2(x-x_i)l'_i)l_i^2(x), \quad \tilde{h}_i(x) = (x-x_i)l_i^2(x)$$

If $f \in C^{(2n+2)}[a, b]$, $p(x)$ is the Hermite interpolating polynomial, then

$$f(x) = p(x) + \frac{f^{(2n+2)}(\xi)}{(2n+2)!} (x-x_0)^2 \dots (x-x_n)^2$$

Splines

Let $a = x_0 < x_1 < \dots < x_{n-1} < x_n = b$. A spline of degree m is a function $S(x)$ which satisfies the following conditions:

- 1) For $x \in [x_i, x_{i+1}]$, $S(x) = S_i(x)$: polynomial of degree $\leq m$
- 2) $S^{(m-1)}(x)$ exists and is continuous at the interior points x_1, \dots, x_{n-1} , i.e. $\lim_{x \rightarrow x_i^-} S_{i-1}^{(m-1)}(x) = \lim_{x \rightarrow x_i^+} S_i^{(m-1)}(x)$

Let f be defined on $[a, b]$, $a = x_0 < x_1 < \dots < x_{n-1} < x_n = b$ and let S be the natural cubic spline interpolant of f . Then

$$1) |f(x) - S(x)| \leq \frac{5}{384} \max_{a \leq x \leq b} |f^{(4)}(x)| h^4$$

where $h = \max_i |x_{i+1} - x_i|$

$$\int_a^b (S''(x))^2 dx \leq \int_a^b (f''(x))^2 dx$$

Numerical Integration

$$\int_a^b f(x) dx \sim \sum_{i=0}^n c_i f(x_i)$$

Trapezoid Rule:

$$T(h) = h \left(\frac{1}{2} f(x_0) + f(x_1) + \dots + f(x_{n-1}) + \frac{1}{2} f(x_n) \right)$$

Local Error Estimate:

$$\int_a^{a+h} f(x) dx = h \frac{f(a) + f(a+h)}{2} - \frac{h^3}{12} f''(\xi)$$

Global Error Estimate:

$$\int_a^b f(x) dx = T(h) - \frac{f''(\xi)}{12} h^2 (b-a)$$

Simpson's Rule:

$$S(h) = h \left(\frac{1}{3} f(x_0) + \frac{4}{3} f(x_1) + \frac{2}{3} f(x_2) + \dots \right. \\ \left. \dots + \frac{2}{3} f(x_{n-2}) + \frac{4}{3} f(x_{n-1}) + \frac{1}{3} f(x_n) \right)$$

Error:

$$\int_a^b f(x) dx = S(h) - \frac{f^{(4)}(\xi)}{180} h^4 (b-a)$$

Orthogonal Polynomials:

The *inner product* of two functions f and g on $[a, b]$ with the weighting function $w(x)$ is

$$\langle f, g \rangle = \int_a^b f(x)g(x)w(x)dx$$

Properties:

- 1) $\langle f, f \rangle \geq 0$, $\langle f, f \rangle = \|f\|^2 = 0 \Leftrightarrow f = 0$
- 2) $\langle f, \alpha g + h \rangle = \alpha \langle f, g \rangle + \langle f, h \rangle$

Gaussian Quadrature:

$$\int_{-1}^1 f(x) dx \sim \sum_{i=1}^n c_i f(x_i)$$

where $x_i, i = 1, \dots, n$ are roots of Legendre polynomial $P_n(x)$.

IVP for ODEs:

$$y' = f(t, y), \quad y(a) = \alpha, \quad a \leq t \leq b$$

Euler's Method:

$$u_{n+1} = u_n + hf(t_n, u_n), \quad u_0 = \alpha$$

Local Truncation Error: $\tau_n = \frac{h^2}{2} y''(\tilde{t}_n)$

Global Error:

$$|y_n - u_n| \leq \frac{hM}{2L} (e^{L(t_n-a)} - 1)$$

where L is a Lipschitz constant, $M = \max |y''(t)|$.

Modified Euler's Method:

$$k_1 = f(t_n, u_n), \quad k_2 = f(t_n + h, u_n + hk_1)$$

$$u_{n+1} = u_n + \frac{h}{2}(k_1 + k_2)$$

4th Order Runge-Kutta Method:

$$k_1 = f(t_n, u_n), \quad k_2 = f(t_n + \frac{h}{2}, u_n + \frac{h}{2}k_1)$$

$$k_3 = f(t_n + \frac{h}{2}, u_n + \frac{h}{2}k_2), \quad k_4 = f(t_n + h, u_n + hk_3)$$

$$u_{n+1} = u_n + \frac{h}{6}(k_1 + 2k_2 + 2k_3 + k_4)$$

Backward Euler's Method:

$$u_{n+1} = u_n + hf(u_{n+1})$$

System of ODEs:

$$y' = Ay, \quad A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

Exact Solution:

$$y(t) = \alpha_1(0)e^{\lambda_1 t} p_1 + \alpha_2(0)e^{\lambda_2 t} p_2$$

Forward Euler:

$$u_n = \alpha_1(0)(1 + h\lambda_1)^n p_1 + \alpha_2(0)(1 + h\lambda_2)^n p_2$$

Backward Euler:

$$u_n = \alpha_1(0) \left(\frac{1}{1 - h\lambda_1} \right)^n p_1 + \alpha_2(0) \left(\frac{1}{1 - h\lambda_2} \right)^n p_2$$

Multistep Methods:

General 2-Step Method:

$$\alpha_0 u_{n+1} + \alpha_1 u_n + \alpha_2 u_{n-1} = h[\beta_0 f(u_{n+1}) + \beta_1 f(u_n) + \beta_2 f(u_{n-1})]$$

Adams-Bashforth:

$$u_{n+1} = u_n + \frac{h}{2}[3f(u_n) - f(u_{n-1})]$$

Adams-Moulton

$$u_{n+1} = u_n + \frac{h}{12}[5f(u_{n+1}) + 8f(u_n) - f(u_{n-1})]$$

Leap-Frog

$$u_{n+1} = u_{n-1} + 2hf(u_n)$$

BDF: Backward Differentiation Formula — Gear's Method

$$\frac{3}{2}u_{n+1} - 2u_n + \frac{1}{2}u_{n-1} = hf(u_n)$$

Computing eigenvalues and eigenvectors

$$Ax = \lambda x, \quad x \neq 0$$

λ : eigenvalue

x : associated eigenvector

Power method:

Idea: v, Av, A^2v, \dots

1. $v^{(0)}$: given, $\|v^{(0)}\|_2 = 1$
2. for $k = 1, 2, \dots$
3. $w = Av^{(k-1)}$
4. $v^{(k)} = w/\|w\|_2$
5. $\lambda^{(k)} = (v^{(k)})^T Av^{(k)}$

Inverse iteration:

Idea: apply power method to $A^{-1}, (A - \mu I)^{-1}, \mu$: shift

1. $v^{(0)}$: given, $\|v^{(0)}\|_2 = 1$
2. for $k = 1, 2, \dots$
3. solve $(A - \mu I)w = v^{(k-1)}$
4. $v^{(k)} = w/\|w\|_2$
5. $\lambda^{(k)} = (v^{(k)})^T Av^{(k)}$

Rayleigh quotient iteration:

Idea: update μ

1. $v^{(0)}$: given, $\|v^{(0)}\|_2 = 1, \lambda^{(0)} = (v^{(0)})^T Av^{(0)}$
2. for $k = 1, 2, \dots$
3. solve $(A - \lambda^{(k-1)}I)w = v^{(k-1)}$
4. $v^{(k)} = w/\|w\|_2$
5. $\lambda^{(k)} = (v^{(k)})^T Av^{(k)}$

Least Squares:

$A\vec{z} = \vec{b}$: $m \times n$ system with $m \geq n$

$A^T A\vec{z} = A^T \vec{b}$: normal equations