# CS 371

## Condition of a problem

$$\kappa_{A} = \frac{||\Delta z||}{||\Delta x||}$$

$$\kappa_{R} = \frac{\left(\frac{||\Delta z||}{||z||}\right)}{\left(\frac{||\Delta x||}{||x||}\right)}$$

$$\Delta x = x - \hat{x} \ \delta x = \frac{|\Delta x|}{x}$$

$$||x||_1 = \sum_{i=1}^n |x_i| \quad ||x||_p = \sqrt[p]{\sum_{i=1}^n x_i^p}$$

$$||x||_{\infty} = \max_{1 \le i \le n} |x_i|$$

 $|x \cdot y| < ||x|| \times ||y||$  (Cauchy-Schwartz Inequality.)

$$||x||_{\infty} \le ||x||_2 \le ||x||_1 \le n||x||_{\infty}$$

$$||x||_{\infty} \le ||x||_p \le n||x||_{\infty}$$

$$e^x = \sum_{n=1}^{\infty} \frac{x^n}{n!}$$

$$\cos(x) = \sum_{n=1}^{\infty} \frac{(-1)^n}{(2n)!} x^{2n} = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \dots$$

$$\sin(x) = \sum_{n=1}^{\infty} \sum_{n=1}^{\infty} \frac{(-1)^n}{(2n+1)!} x^{2n+1} = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \dots$$

# Numerical Linear Algebra

$$\underbrace{A}_{n \times p} \stackrel{p \times m}{B} \qquad \text{flops} = nm(p+p-1)$$

#### LU Factorization

$$PA = \underbrace{LU}_{\text{Lower and upper triangular matrix}}$$

$$\mathbf{A} \Rightarrow \mathbf{0}$$

$$\mathbf{I_n} \underset{\text{same row operations}}{\Rightarrow} L' \underset{\text{flip off-diagonal elements}}{\Rightarrow} L$$

$$L\vec{\mathbf{y}} = \vec{\mathbf{b}} \Rightarrow U\vec{\mathbf{x}} = \vec{\mathbf{y}}$$
forward sub back sub

## Computational cost

**Decomposing:**  $\frac{2n^3}{3} + O(n^2)$ 

Forward and backward sub:  $n^2 + O(n)$ 

### Determinants

- 1.  $\det(BC) = \det(B) \det(C)$
- 2. U upper triangular  $\Rightarrow \det U = \prod_{i=1}^n u_{ii}$
- 3. L lower triangular  $\Rightarrow$  det  $U = \prod_{i=1}^{n} u_{ii}$ 4. P Permutation matrix  $\Rightarrow$  det  $P_{\text{even}} = +1$ , det  $P_{\text{odd}} = -1$
- 5. If det  $\mathbf{A} \neq 0$  then  $\mathbf{A}x = b$  has unique solution
- 6. If det  $\mathbf{A} = 0$  then  $\mathbf{A}x = b$  has 0 or infinite solutions.

### Condition and Stability

$$\left\|\mathbf{A}\right\|_{p} = \max_{\left\|\vec{\mathbf{x}}\right\|_{p} \neq 0} \frac{\left\|\mathbf{A}\vec{\mathbf{x}}\right\|_{p}}{\left\|\vec{\mathbf{x}}\right\|_{p}}$$

 $\|\mathbf{A}\|_1 = \text{maximum abs column sum}$ 

 $\|\mathbf{A}\|_{\infty} = \max \max \text{maximum abs row sum}$ 

$$\|A\|_p = 0 \iff \|A\| = 0, \ \|cA\|_p = |c| \|A\|_p,$$

$$||A + B||_p \le ||A||_p + ||B||_p$$

$$\kappa_p(A) = ||A||_p ||A^{-1}||_p$$
 (Condition number)

$$\kappa_2(A) = \sqrt{\frac{\lambda_{\max}(A^T A)}{\lambda_{\min}(A^T A)}}$$

$$\kappa(B) = \infty$$
 if B is singular

$$\frac{1}{\kappa(B)}$$
 how close  $B$  is to a singular matrix

$$\vec{\mathbf{r}} = \vec{\mathbf{b}} - A\vec{\mathbf{u}}$$
 (residual)

# Root finding

**Bisection Method** make intervals  $[a_k, b_k], f(a_0)f(b_0) \leq 0$ 

$$a_k = \begin{cases} a_{k-1} & f(\frac{a_{k-1} + b_{k-1}}{2}) \cdot f(a_{k-1}) \le 0\\ \frac{a_{k-1} + b_{k-1}}{2} & \text{otherwise} \end{cases}$$

$$b_k = \begin{cases} b_{k-1} & f(\frac{a_{k-1} + b_{k-1}}{2}) \cdot f(a_{k-1}) > 0 \\ \frac{a_{k-1} + b_{k-1}}{2} & \text{otherwise} \end{cases}$$

# of steps for tolerance  $t \geq \frac{|b_0 - a_0|}{2N}$ Newton's Method

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$$

#### Secant Method

$$x_{k+1} = x_k - f(x_k) \left[ \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} \right]$$

## Rate of Convergence

Error:  $e_k = x_k - x^*$ 

Order of convergence:  $x_k$  converges with order q if and only if  $x_i \to x^*$ 

$$\lim c_i = N \in [0, \infty)$$

$$|e_{i+1}| = c_i |e_i|^q$$
 Bisection Method  $q = 1$   
Secant Method  $q = \frac{1+\sqrt{5}}{2}$   
Newton's Method  $q = 2$ 

#### Fourier Series

$$\omega = 2\pi f$$

$$g_n(x) = \frac{a_0}{2} + \sum_{k=0}^{n} \left[ a_k \cos\left(k\frac{2\pi}{b-a}\right) + b_k \sin\left(k\frac{2\pi}{b-a}\right) \right]$$

$$a_k = \frac{2}{b-a} \int_a^b f(x) \cos\left(k \frac{2\pi x}{b-a}\right)$$

$$b_k = \frac{2}{b-a} \int_a^b f(x) \sin\left(k \frac{2\pi x}{b-a}\right)$$

$$c_k = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(t) \exp(ikt) dt \ [a, b] = [-\pi, \pi]$$

$$c_k = \frac{1}{2} \left( a_k + i b_k \right)$$

$$f(x) = f(-x)$$
 Even

$$f(-x) = -f(x)$$
 Odd

$$f(t)$$
 even  $\Rightarrow b_k = 0 \ \forall k \ (\cos(x) = \cos(-x))$ 

$$f(t) \text{ odd } \Rightarrow a_k = 0 \ \forall k \ (\sin(-x) = -\sin(x))$$

$$V = \left\{ f(x) : \sqrt{\int_a^b |f(x)|^2 \, \mathrm{d}x} < \infty \right\}$$

If  $f(x) \in V$ , Fourier series of  $g_n(x) \to f(x)$  on [a, b]

$$\bar{c_k} = c_{-k}$$

$$a_{-k} = a_k \ b_{-k} = -b_{-k}$$

$$a_k = 2\operatorname{Re}(c_k) , b_k = -2\operatorname{Im}(c_k) , b_0 = 0 , c_0 = \frac{1}{2}a_0$$

# Roots of unity

$$W_N = \exp\left(\frac{2\pi i}{N}\right) n$$
—th root of unity

$$W_N^k = \exp\left(\frac{2k\pi}{N}\right)$$

$$(W_N^k)^N = 1$$

$$(W_N)^{-k} = W_N^{N-k}$$

### DFT

$$f[i] = f(i)$$
 for  $0 \le i \le N - 1$  (Sampled for N points)

$$F[k] = DFT\{f[n]\} = \frac{1}{N} \sum_{n=0}^{N-1} f[n] W_N^{-kn}$$

$$f[n] = IDFT\{F[k]\} = \sum_{k=0}^{N-1} F[k]W_N^{kn}$$

$$\vec{F} = W\vec{f}$$

$$W_{ij} = \frac{1}{N} W_N^{-ij}$$
 for  $0 \le i, j \le N - 1$ 

$$W_{ij}^{-1} = W_N^{ij} \text{ for } 0 \le i, j \le N - 1 \quad W_{ij}^{-1} = N(\overline{W_{ij}})$$

Properties of F

$$F[k] = F[k + sN]$$
 for  $s \in \mathbb{Z}$ 

$$\overline{F[k]} = F[-k]$$

 $Re\{F[k]\}$  is even in k

 $\operatorname{Im}\{F[k]\}\ \text{is odd in } k$ 

$$f[n] = f[N-n] = f[-n] f$$
 is even in  $n$ 

$$f[n] = -f[N-n] = -f[-n]$$
 f is odd in n

$$f[n]$$
 is even in  $n \Rightarrow \operatorname{Im}(F[k]) = 0$  (DFT is real)

f[n] is odd in  $n \Rightarrow \text{Re}(F[k]) = 0$  (DFT is purely imaginary)

# Aliasing and the Sample Theorem

 $f_s = \frac{N}{T}$  Sampling Rate

Sampling Theorem: If a function f(t) is bandwidth limited to frequencies smaller than  $f_c$  (max frequency  $\leq f_c$ ) and f(t) is sampled at a rate  $f_s \geq 2f_c$  then the function is completely determined by its samples f[n].

**Sampling Theorem(ii)**: For a fixed sample f[n] with a fixed sampling rate  $f_s$ . Then if the maximum frequency of a signal  $f_c \leq \frac{f_s}{2}$  that can be deconstructed from f[n] such that is free of aliasing errors (DFT if free of aliasing errors).

### Fast Fourier Transform

flops 
$$F[k] = 2N \Rightarrow$$
 flops  $\vec{F} = 2N^2$  (Using sums)

$$E[k]$$
 (DFT of even indexed part)

$$O[k]$$
 (DFT of odd indexed part)

$$F[k] = \frac{1}{2} \left( E[k] + W_N^{-k} O[k] \right) \text{ for } k = 0, \dots, \frac{N}{2} - 1$$
$$F[k + \frac{N}{2}] = \frac{1}{2} \left( E[k] - W_N^{-k} O[k] \right) \text{ for } k = 0, \dots, \frac{N}{2} - 1$$

flops 
$$ec{F} = rac{5}{2} N \log_2(N)$$

## Power Spectrum and Parseval's Theorem

Power spectrum of f(t) or f[n] is  $|F[k]|^2$ Parseval's Theorem (Continous)

$$\frac{1}{b-a} \int_a^b f(t)^2 dt = \sum_{k=-\infty}^{\infty} |c_k|^2$$

### Parseval's Theorem (Discrete)

$$\frac{1}{N} \sum_{k=0}^{N-1} |f[n]|^2 = \sum_{k=0}^{N-1} |F[k]|^2$$

# Interpolation

For a basis  $B = {\{\phi_j(x)\}_{j=0}^m \text{ interpolating points } \{(x_i, f_i)\}_{i=0}^n}$ 

$$y(x) = \sum_{i=0}^{m} a_i \phi_i(x)$$

$$\Phi = \begin{pmatrix} \phi_0(x_0) & \phi_1(x_0) & \cdots & \phi_m(x_0) \\ \phi_0(x_1) & \phi_1(x_1) & \cdots & \phi_m(x_1) \\ \vdots & \vdots & \cdots & \vdots \\ \phi_0(x_n) & \phi_1(x_n) & \cdots & \phi_m(x_n) \end{pmatrix}$$

 $=\vec{f}$  (Solves for  $a_i$ )

### Vandermonde Polynomial:

 $y(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n$  for (n+1) points  $(x_i, f_i)$ .

$$\Phi = V = \begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_n \end{pmatrix}$$

$$\det(V) = \prod_{0 \le i < j \le n} (x_j - x_i)$$

Computing V takes  $\Theta(N^3)$  flops

Computing y(x) takes 3N flops

 ${\bf Lagrange\ Interpolation\ The\ Lagrange\ basis\ functions\ are\ defined\ as}$ 

$$\ell_i(x_j) = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

$$\ell_i(x) = \prod_{j=0, j \neq i}^n \frac{x - x_j}{x_i - x_j}$$
 (basis)

 $\Phi = L = I_{n \times n}$ 

Computing I takes 0 flops

Computing pre-computed terms takes  $2N^2 + 2N$  flops

Computing y(x) takes 5N flops

#### Newton Interpolation

$$\pi_j(x) = \prod_{i=0}^{j-1} (x - x_i) \ \pi_0(x) = 1$$
 (basis)

$$\pi_j(x_i) = 0 \text{ if } i < j$$

$$\Pi = \begin{pmatrix} \pi_0(x_0) & \cdots & \mathbf{0} \\ \vdots & \ddots & \vdots \\ \pi_n(x_n) & \cdots & \pi_n(x_n) \end{pmatrix}$$
 (II is lower triangular)

Computing  $\Pi$  takes  $\Theta(N^2)$  flops

Computing pre-computed terms takes  $\Theta(N^2)$  flops

Computing y(x) takes  $\Theta(N)$  flops

Extending newton polynomial with additional point  $(x_{n+1}, f_{n+1})$ 

$$y_{n+1}(x) = y_n(x) + a_{n+1}\pi_{n+1}(x)$$
  $a_{n+1} = \frac{f_{n+1} - y_n(x_{n+1})}{\pi_{n+1}(x)}$ 

Hermite Interpolation Given  $\{(x_i, f_i, f'_i)\}_{i=0}^n$  interpolating polynomial has degree 2n+1.

Chebyshev Points on interval [a, b] the n + 1 points are

$$x_j = \frac{a+b}{2} + \frac{b-a}{2} \cos\left(\frac{2j+1}{2n+2}\pi\right) \text{ for } j = 0,\dots,n$$

Cubic Splines Smoothness conditions:  $y_j'(x_j) = y_{j+1}'(x_j)$  and  $y_j''(x_j) = y_{j+1}''(x_j)$  for cubic splines free boundary:  $y_1''(x_0) = 0$ ,  $y_n''(x_n) = 0$ , clamped boundary:  $y_1''(x_0) = f_0'$ ,  $y_n''(x_n) = f_n'$ , periodic boundary if  $f_0 = f_n$ :  $y_1'(x_0) = y_n'(x_n)$  and  $y_1''(x_0) = y_n''(x_n)$ . Cubic splines can be solved in  $\Theta(N)$  flops

### Regression

 $\Phi$  with fewer basis function m+1 < n+1. the system is over-determined we can find solution  $\vec{a}$  such that the residue  $\min \left\| \Phi \vec{a} - \vec{f} \right\|_2^2$  is minimized.

$$\underbrace{\Phi}_{(n+1)\times(m+1)}\underbrace{\vec{a}}_{(m+1)\times 1} + \vec{r} = \vec{f}$$

$$\boxed{\Phi^T \Phi \vec{a} = \Phi^T \vec{f}}$$
 (normal equation

$$\begin{pmatrix} a \\ b \end{pmatrix} = \frac{1}{(n+1)\sum x_i^2 - \left(\sum x_i\right)^2} \bigg( \frac{\left(\sum x_i^2\right)(\sum f_i) - (\sum x_i)(\sum x_i f_i)}{(n+1)(\sum x_i f_i) - (\sum x_i)(\sum f_i)} \bigg)$$

## Numerical Integration

$$I_M = \sum_{i=1}^{n} (x_i - x_{i-1}) f\left(\frac{x_i - x_{i-1}}{2}\right)$$
 (Midpoint rule)

(basis) 
$$I_T = \sum_{i=1}^{n} \frac{(x_i - x_{i-1})}{2} (f(x_{i-1}) + f(x_i))$$
 (Trapezoid Rule)

$$I_S = \sum_{i=1}^{n} \frac{x_i - x_{i-1}}{6} \left[ f(x_{i-1}) + 4f(x_{i-\frac{1}{2}}) + f(x_i) \right]$$

Where 
$$x_{i-\frac{1}{2}} = \frac{x_{i-1} + x_i}{2}$$
 (Simpsons Rule)

#### **Error Bounds**

$$|E_M| \le \frac{(b-a)^3}{24n^2} \max_{x \in [a,b]} |f''(x)|$$
 (Midpoint rule error bound)

$$|E_T| \leq \frac{(b-a)^3}{12n^2} \max_{x \in [a,b]} \left| f''(x) \right| \qquad \text{(Trapezoid rule error bound)}$$

$$|E_S| \le \frac{(b-a)^5}{2880n^4} \max_{x \in [a,b]} \left| f^{(4)}(x) \right|$$
 (Simpsons rule error bound)

**Degree of precision**:  $\hat{I}$  has degree of precision m if  $E = I - \hat{I} = 0$  for polynomial p(x) such that  $\deg p \leq m$ . Or  $\hat{I}$  integrates any polynomial p with  $\deg p \leq m$  exactly. Midpoint and Trapezium rule:  $\deg_p = 1$ , Simpsons rule:  $\deg_p = 3$ .