

# CS 450: Numerical Analysis

## Lecture 21

### Chapter 7 Numerical Integration and Differentiation

#### Basic Numerical Quadrature Methods

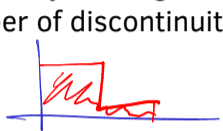
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# Integrability and Sensitivity

- ▶ Function  $f$  is integrable if continuous and bounded, in practice a finite number of discontinuities is also ok:



$$I(f) = \int_a^b f(x) dx, \quad \|f\|_{\infty} = \max_{x \in [a, b]} |f(x)|$$

- ▶ The condition number of integration is bounded by the distance  $b - a$ :

$$\hat{f} = f + \delta f, \quad \|\hat{f} - f\|_{\infty} \leq \|\delta f\|_{\infty}$$

$$|I(\hat{f}) - I(f)| = |I(\hat{f} - f)| = |I(\delta f)|$$

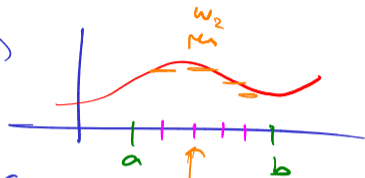
$$\leq (b-a) \|\delta f\|_{\infty}$$

# Quadrature Rules

- ▶ To approximate the integral  $I(f)$ , compute a weighted sum of points:

$$I(f) \approx Q_n(f) = \sum_{i=1}^n w_i f(x_i)$$

$\uparrow$   
 $n$ -points



$w_i$  - quadrature weights  
 $x_i$  - nodes

so if  $Q_n$  uses  $x_2$  a subset of nodes

- ▶ For a fixed set of  $n$  nodes, unique quadrature weights give exact  $(n-1)$ -degree quadrature rule:

polynomial interpolant of degree  $n-1$  is unique, and so its integral is unique.

If  $y_i = f(x_i)$ ,  $P_{n-1}(x) = \sum_{i=1}^n y_i \ell_i(x)$ , so  $Q_n(f) = \sum_{i=1}^n y_i \int_a^b \ell_i(x) dx$

$\uparrow$   
 Lagrange basis function

rule is progressive of  $Q_{n+1}$

## Quadrature Rules and Error

- ▶ Quadrature weights can be alternatively determined for a rule by solving the moment equations:

$$\begin{bmatrix} \ell_1(x_1) & \dots & \ell_1(x_n) \\ \vdots & & \vdots \\ \ell_n(x_1) & \dots & \ell_n(x_n) \end{bmatrix} \begin{bmatrix} w \\ \vdots \\ w \end{bmatrix} = \begin{bmatrix} I(\ell_1) \\ \vdots \\ I(\ell_n) \end{bmatrix}$$

- ▶ We can approximate the error bound for a polynomial quadrature rule by

$$\begin{aligned} |I(f) - Q_n(f)| &= |I(f - p_{n-1})| \\ &= |I(\alpha_1 x^n + \alpha_2 x^{n+1} + \dots)| \\ &= \frac{(b-a)}{4n!} \|f^{(n)}\|_\infty \cdot h^n, \quad h = \max_i (x_{i+1} - x_i) \end{aligned}$$

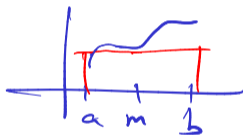
# Newton-Cotes Quadrature

- ▶ **Newton-Cotes** quadrature rules are defined by equispaced nodes on  $[a, b]$ :

closed if including  $a, b$ :  $x_i = a + \frac{(i-1)(b-a)}{n-1}$   
open otherwise

- ▶ The **midpoint rule** is the  $n = 1$  open Newton-Cotes rule:  $x_i = a + \frac{i(b-a)}{n+1}$

$$x_1 = a + \frac{(b-a)}{2} = \frac{a}{2} + \frac{b}{2}, \quad f\left(\frac{a+b}{2}\right)$$



- ▶ The **trapezoid rule** is the  $n = 2$  closed Newton-Cotes rule:

$$\frac{f(a) + f(b)}{2}$$



- ▶ **Simpson's rule** is the  $n = 3$  closed Newton-Cotes rule:

quadratic interpolant

## Error in Newton-Cotes Quadrature

- Consider the Taylor expansion of  $f$  about the midpoint of the integration interval  $m = (a + b)/2$ :

$$f(x) = f(m) + f'(m)(x - m) + \frac{f''(m)}{2}(x - m)^2 + \dots$$

$$\int(f) = \underbrace{(b-a)f(m)}_{M(f)} + 0 + \underbrace{\frac{f''(m)}{24}(b-a)^3}_{E(f)} + \dots$$

Error of Midpoint rule

$$\approx E(f) + o((b-a)^5)$$

Trapezoid rule has nearly same bound

$$M(f) - T(f) = \underline{3E(f)} + o((b-a)^5)$$

## Conditioning of Newton-Cotes Quadrature

- ▶ We can ascertain stability of quadrature rules, by considering the amplification of a perturbation  $\hat{f} = f + \delta f$ :

$$\begin{aligned} |Q_n(\hat{f}) - Q_n(f)| &= |Q_n(\delta f)| \\ &= \left| \sum_{i=1}^n w_i \delta f(x_i) \right| \\ &\leq \|w\|_1 \|\delta f\|_\infty \end{aligned}$$

$$\sum_{i=1}^n w_i = 1$$

- ▶ Newton-Cotes quadrature rules have at least one negative weight for any  $n \geq 11$ :

as  $n \rightarrow \infty$

$\|w\|_1 \rightarrow \infty$  dep. on  $\sum_{i=1}^n w_i = 1$

# Clenshaw-Curtis Quadrature

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- ▶ To obtain a more stable quadrature rule, we need to ensure the integrated interpolant is well-behaved as  $n$  increases:

Chebyshev nodes for quadrature  
rules defined e.g. by integrate legendre

good stability  
also efficient by use of cos-transform









