

# ⑧ Int and Diff.

## ⑧.1 Num. Integration

$a, b, f$  approximate  $\int_a^b f(x) dx$

Existence, uniqueness:  $f$  integrable  $\Rightarrow \int f$  exists, unique  
or continuous, bounded  $\Rightarrow f$  integrable

## Conditioning

$$\hat{f}(x) := f(x) + e(x)$$

$$\left| \int_a^b f(x) dx - \int_a^b \hat{f}(x) dx \right|$$

$$= \left| \int_a^b e(x) dx \right|$$

$$\leq \int_a^b |e(x)| dx$$

$$\leq (b-a) \underbrace{\max_{x \in [a,b]} |e|}_{\text{abs. condition number}}$$

abs. condition number

## 8.2 Quadrature methods

$$\int_a^b f(x) dx = \sum_{i=1}^n w_i f(x_i)$$

↑                    ↖  
weights            nodes

Idea 1: "interpolatory quadrature"

$$f(x) \approx \sum_{i=1}^n f(x_i) l_i(x)$$

↑  
Lagrange basis

$$\int_a^b f(x) dx \approx \sum_{i=1}^n f(x_i) \int_a^b l_i(x) dx$$

↑  
trickier.

① w/ monomials and (mostly) equispaced points

"Newton-Cotes quadrature"

② w/ Chebyshev polys at Chebyshev nodes

"Clenshaw-Curtis quadrature"

## Method of undetermined coefficients

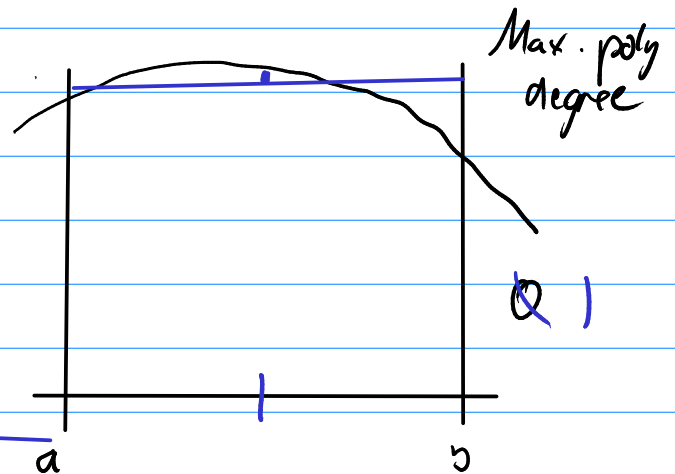
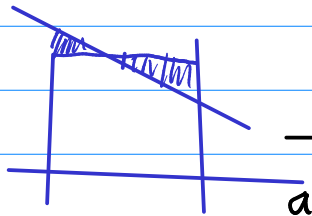
$$b-a = \int_a^b 1 dx = w_1 \cdot 1 + w_2 \cdot 1 + \dots + w_n \cdot 1$$

$$\frac{1}{z} \left( b^{\frac{k+1}{z}} - a^{\frac{k+1}{z}} \right) = \int_a^b x^k dx = w_1 \cdot x_1^k + w_2 \cdot x_2^k + \dots + w_n \cdot x_n^k$$

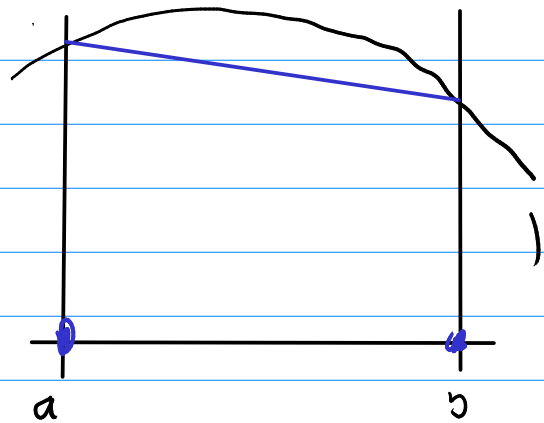
$k+1$

## Examples

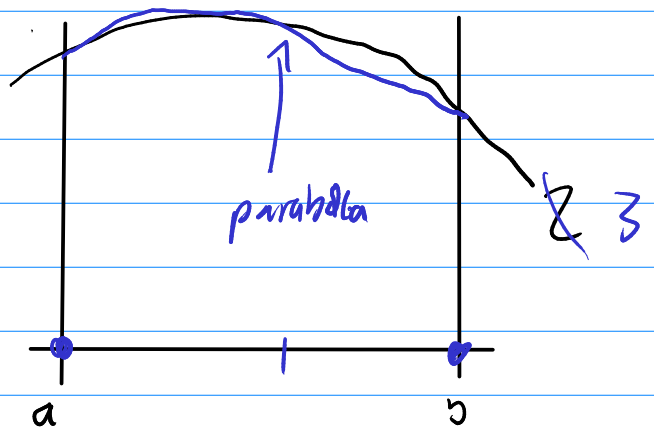
### Midpoint rule



### Trapezoidal rule



### Simpson's rule



Idea: Can estimate error by computing multiple quadratures (such as midpoint + trapez.)

## 8.2 Accuracy and stability (For Newton-Cotes)

Accuracy: Let  $p_n$  be an interpolant of  $f$   
at nodes  $x_1, \dots, x_n \leftarrow \text{deg. } n-1$ .

Recall:  $\sum_{i=1}^n \omega_i f(x_i) = \int_a^b p_n(x)$

$$\begin{aligned} & \left| \int_a^b f(x) dx - \int_a^b p_n(x) dx \right| \\ & \leq \int_a^b |f(x) - p_n(x)| dx \\ & \leq (b-a) \|f - p_n\|_{\infty} \\ \text{interp. err. est.} & \leq \frac{(b-a)}{4} \frac{h^n}{n} \cdot \|f^{(n)}\|_{\infty} \\ & \leq \frac{h^{n+1}}{4} \|f^{(n)}\|_{\infty} \end{aligned}$$

Stability:  $\hat{f}(x) = f(x) + e(x)$

$$\left| \sum_i w_i f(x_i) - \sum_i w_i \hat{f}(x_i) \right|$$

$$= \left| \sum_i w_i e(x_i) \right|$$

$$\leq \sum_i |w_i| \underbrace{|e(x_i)|}_{\leq \|e\|_\infty}$$

$$\leq \left( \sum_i |w_i| \right) \cdot \|e\|_\infty$$

Oscillatory weights  $\rightarrow$  bad quad. rule

WS 26 p1