Chapter 5: Numerical Differentiation and Integration

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Motivation

Given a function f(x),

• Differentiation Problem: compute the derivative of f(x) at a value x_0 ?



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Given a function f(x),

- Differentiation Problem: compute the derivative of f(x) at a value x_0 ?
- Integration Problem: compute the definite integral of f(x) on an interval [a,b]?





Motivation Numerical differentiation Numerical Integration Conclusions

Outline

- Motivation
- Numerical differentiation
 - Finite difference formulas
 - Extrapolation
- 3 Numerical Integration
 - Trapezoid rule
 - Simpson's rule
 - Composite Newton-Cotes formulas
 - Open Newton-Cotes Methods
 - Romberg Integration
 - Adaptive Quadrature
 - Gaussian Quadrature
- 4 Conclusions



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Finite difference formulas

Definition (Derivative)

The derivative of f(x) at a value x is

$$f(x) = \lim_{h \to 0} \frac{f(x+h) - f(x)}{h}.$$



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Theorem (Taylor's Theorem)

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2}f''(c),$$

where x < c < x + h.



Two-point forward-difference formula

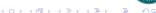
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$$f(x) = \frac{f(x+h) - f(x)}{h} - \frac{h}{2}f'(c),$$

where x < c < x + h.

When h is small, we can approximate f(x) as follows:

$$f(x) = \frac{f(x+h) - f(x)}{h},$$

where $\left|\frac{h}{2}f''(c)\right|$ is the error.



Motivation

Example

• Use the two-point difference formula with h=0.1 to approximate the derivative of $f(x) = \frac{1}{x}$ at x = 2.



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- $f(x) \approx \frac{f(x+h)-f(x)}{h} = \frac{\frac{1}{2 \cdot 1} \frac{1}{2}}{0 \cdot 1} \approx -0.2381$





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Numerical Integration

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- The actual error is |-0.2381-(-0.25)|=0.0119.



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- The actual error is |-0.2381 (-0.25)| = 0.0119.
- The estimation error $\max\{\frac{hf''(c)}{2}\mid 2\leq c\leq 2.1\}$ where $f''(x)=2x^{-3}$.
- \bullet $\frac{0.1f''(c)}{2} = 0.0125 < 0.0119$ where c = 2.



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Three-point centered-difference formula

Theorem (Taylor's Theorem)

Suppose that f(x) is 3-times continuously differentiable.

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \frac{h^3}{6}f''(c_1),$$

$$f(x-h) = f(x) - hf'(x) + \frac{h^2}{2}f''(x) - \frac{h^3}{6}f''(c_2),$$

where $x - h < c_2 < x < c_1 < x + h$.



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Subtracting the above two equations gives the following

$$f'(x) = \frac{f(x+h) - f(x-h)}{2h} - \frac{h^2}{6}f''(c_1) - \frac{h^2}{6}f''(c_2).$$



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Could we combine the last two terms into one? Yes!



Theorem (Generalized intermediate value theorem)

Let f be a continuous function on the interval [a,b]. Let x_1, \dots, x_n be points in [a,b], and $a_1, \dots, a_n > 0$. Then, there exists a number c between a and b s.t.

$$(a_1 + \cdots + a_n) f(c) = a_1 f(x_1) + \cdots + a_n f(x_n).$$



Proof.

Let $f(x_i)$ and $f(x_j)$ be the minimum and maximum of $f(x_1), \dots, f(x_n)$ respectively.





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So
$$f(x_i) \leq \frac{a_1 f(x_1) + \dots + a_n f(x_n)}{a_1 + \dots + a_n} \leq f(x_j)$$
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By the Intermediate Value Theorem, there is a number c between x_i and x_j s.t.

$$f(c) = \frac{a_1 f(x_1) + \dots + a_n f(x_n)}{a_1 + \dots + a_n}.$$

Moreover, a < c < b.



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- The actual error is |-0.2506 (-0.25)| = 0.0006.
- The estimation error $\max\{\frac{h^2f'''(c)}{6}\mid 1.9\leq c\leq 2.1\}$ where $f'''(x)=-6x^{-4}.$
- \bullet $\frac{(0.2)^2 f'''(c)}{6} = 0.0031 > 0.0006$ where c = 1.9.



Theorem (Taylor's Theorem)

Suppose that f(x) is 4-times continuously differentiable.

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \frac{h^3}{6}f'''(x) + \frac{h^4}{24}f^{(4)}(c_1),$$

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Adding the above two equations gives the following

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Combing the last two terms leads to

$$f(x+h) + f(x-h) - 2f(x) = h^2 f'(x) + \frac{h^4}{12} f^{(4)}(c),$$



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$$f''(x) = \frac{f(x-h) - 2f(x) + f(x+h)}{h^2} - \frac{h^2}{12}f^{(4)}(c),$$

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Definition (Three-point centered-difference formula for second derivative)

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- The correct derivative $f'(x) = 2x^{-3} = 0.25$.
- The actual error is |0.2506 0.25| = 0.0006.
- The estimation error $\max\{\frac{h^2}{12}f^{(4)}(c) \mid 1.9 \le c \le 2.1\}$ where $f^{(4)}(x) = 24x^{-5}$.
- \bullet $\frac{(0.2)^2 f^{(iv)}(c)}{12} = 0.0025 > 0.0006$ where c = 1.9.



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Definition

$$Q = F(h) + K_n h^n + K_{n+1} h^{n+1} + \cdots,$$

where K_n, K_{n+1}, \cdots are constants.

We call F(h) is an order n formula for approximating a given quantity Q.

- The truncation error for F(h) is $O(h^n)$, in general K_nh^n , unless K_{n+1}, K_{n+2}, \cdots is very large.
- $Q F(0.1) \approx 0.1^n \cdot K_n$.
- $Q F(0.01) \approx 0.01^n \cdot K_n$



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Example

Using Taylor expansion, we obtain that

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \frac{h^3}{6}f'''(x) + \frac{h^4}{24}f^{(4)}(x) + \frac{h^5}{120}f^{(5)}(x) + \cdots,$$

$$f(x-h) = f(x) - hf'(x) + \frac{h^2}{2}f''(x) - \frac{h^3}{6}f'''(x) + \frac{h^4}{24}f^{(4)}(x) - \frac{h^5}{120}f^{(5)}(x) + \cdots,$$



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Subtracting the two equations, we get that

$$f'(x) = \frac{f(x+h) - f(x-h)}{2h} - \frac{h^2}{6}f'''(x) - \frac{h^4}{120}f^{(5)}(x) - \cdots$$



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Hence, $\frac{f(x+h)-f(x-h)}{2h}$ is an order 2 formula for approximating f'(x).

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Adding the two equations, we get that

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Conclusions



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- $Q = F_n(h) + \frac{K_n h^n}{h^n} + K_{n+1} h^{n+1} + \cdots;$





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 $F_{n+1}(n) = \frac{2^n F_n(\frac{h}{2}) - F_n(h)}{2^n - 1} \text{ is an order } n+1 \text{ formula for approximating } Q.$



Extrapolated formula for the derivative

Example

- $f'(x) = \frac{f(x+h)-f(x-h)}{2h} \frac{h^2}{6}f'''(x) \frac{h^4}{120}f^{(5)}(x);$
- $F_2(x) = \frac{f(x+h) f(x-h)}{2h}$;
- $\bullet \ \ F_4(x) = \frac{2^2F_2(\frac{h}{2}) F_2(h)}{2^2 1} = \frac{f(x h) 8f(x \frac{h}{2}) + 8f(x + \frac{h}{2}) f(x + h)}{6h}$
- $F_4(x)$ is of order 4 since the order 3 error terms cancel out.





- $f''(x) = \frac{f(x+h)-2f(x)+f(x-h)}{2h} \frac{h^2}{12}f^{(4)}(x) \cdots;$
- $F_2(x) = \frac{f(x+h)-2f(x)+f(x-h)}{2h}$;
- $F_4(x) = \frac{2^2 F_2(\frac{h}{2}) F_2(h)}{2^2 1} = \frac{-f(x h) + 16f(x \frac{h}{2}) 30f(x) + 16f(x + \frac{h}{2}) f(x + h)}{3h^2}$
- $F_4(x)$ is of order 4 since the order 3 error terms cancel out.





Numerical differentiation **Numerical Integration** Conclusions

Outline

Motivation

- - Finite difference formulas
 - Extrapolation
- **Numerical Integration**
 - Trapezoid rule
 - Simpson's rule
 - Composite Newton-Cotes formulas
 - Open Newton-Cotes Methods
 - Romberg Integration
 - Adaptive Quadrature
 - Gaussian Quadrature





- f: a function with a continuous second derivative on $[x_0, x_1]$.
- $y_0 = f(x_0)$ and $y_1 = f(x_1)$.





Motivation

- f: a function with a continuous second derivative on $[x_0, x_1]$.
- $y_0 = f(x_0)$ and $y_1 = f(x_1)$.
- $f(x) = y_0 \frac{x x_1}{x_0 x_1} + y_1 \frac{x x_0}{x_1 x_0} + \frac{(x x_0)(x x_1)}{2!} f'(c) = P(x) + E(x)$
- P(x): the Lagrange interpolating polynomial $y_0 \frac{x-x_1}{x_0-x_1} + y_1 \frac{x-x_0}{x_1-x_0}$;
- E(x): the error formula $\frac{(x-x_0)(x-x_1)}{2!}f''(c)$ where $x_0 \le c \le x_1$.



Conclusions



Motivation

• f: a function with a continuous second derivative on $[x_0, x_1]$.

- $y_0 = f(x_0)$ and $y_1 = f(x_1)$.
- $f(x) = y_0 \frac{x x_1}{x_0 x_1} + y_1 \frac{x x_0}{x_1 x_0} + \frac{(x x_0)(x x_1)}{2!} f'(c) = P(x) + E(x)$
- P(x): the Lagrange interpolating polynomial $y_0 \frac{x-x_1}{x_0-x_1} + y_1 \frac{x-x_0}{x_1-x_0}$;
- E(x): the error formula $\frac{(x-x_0)(x-x_1)}{2!}f''(c)$ where $x_0 \le c \le x_1$.



Conclusions

$$\int_{x_0}^{x_1} P(x) dx = \int_{x_0}^{x_1} y_0 \frac{x - x_1}{x_0 - x_1} dx + \int_{x_0}^{x_1} y_1 \frac{x - x_0}{x_1 - x_0} dx$$



$$\int_{x_0}^{x_1} P(x) dx = \int_{x_0}^{x_1} y_0 \frac{x - x_1}{x_0 - x_1} dx + \int_{x_0}^{x_1} y_1 \frac{x - x_0}{x_1 - x_0} dx$$
$$= y_0 \frac{h}{2} + y_1 \frac{h}{2} \quad (h = x_1 - x_0)$$

Numerical differentiation





$$\int_{x_0}^{x_1} P(x) dx = \int_{x_0}^{x_1} y_0 \frac{x - x_1}{x_0 - x_1} dx + \int_{x_0}^{x_1} y_1 \frac{x - x_0}{x_1 - x_0} dx$$
$$= y_0 \frac{h}{2} + y_1 \frac{h}{2} \quad (h = x_1 - x_0)$$
$$= h \frac{y_0 + y_1}{2}$$



$$\int_{x_0}^{x_1} P(x) dx = \int_{x_0}^{x_1} y_0 \frac{x - x_1}{x_0 - x_1} dx + \int_{x_0}^{x_1} y_1 \frac{x - x_0}{x_1 - x_0} dx$$
$$= y_0 \frac{h}{2} + y_1 \frac{h}{2} \quad (h = x_1 - x_0)$$
$$= h \frac{y_0 + y_1}{2}$$

$$\int_{x_0}^{x_1} E(x) dx = \frac{1}{2!} \int_{x_0}^{x_1} (x - x_0)(x - x_1) f'(c) dx$$



$$\int_{x_0}^{x_1} P(x) dx = \int_{x_0}^{x_1} y_0 \frac{x - x_1}{x_0 - x_1} dx + \int_{x_0}^{x_1} y_1 \frac{x - x_0}{x_1 - x_0} dx$$

$$= y_0 \frac{h}{2} + y_1 \frac{h}{2} \quad (h = x_1 - x_0)$$

$$= h \frac{y_0 + y_1}{2}$$

$$\int_{x_0}^{x_1} E(x) dx = \frac{1}{2!} \int_{x_0}^{x_1} (x - x_0)(x - x_1) f'(c) dx$$
$$= \frac{f'(c)}{2} \int_{x_0}^{x_1} (x - x_0)(x - x_1) dx$$





$$\int_{x_0}^{x_1} P(x) dx = \int_{x_0}^{x_1} y_0 \frac{x - x_1}{x_0 - x_1} dx + \int_{x_0}^{x_1} y_1 \frac{x - x_0}{x_1 - x_0} dx$$
$$= y_0 \frac{h}{2} + y_1 \frac{h}{2} \quad (h = x_1 - x_0)$$
$$= h \frac{y_0 + y_1}{2}$$

$$\int_{x_0}^{x_1} E(x) dx = \frac{1}{2!} \int_{x_0}^{x_1} (x - x_0)(x - x_1) f'(c) dx$$

$$= \frac{f'(c)}{2} \int_{x_0}^{x_1} (x - x_0)(x - x_1) dx$$

$$= \frac{f''(c)}{2} \int_0^h u(u - h) du \quad (h = x_1 - x_0 \text{ and } u = x - x_0)$$



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$$\int_{x_0}^{x_1} P(x) dx = \int_{x_0}^{x_1} y_0 \frac{x - x_1}{x_0 - x_1} dx + \int_{x_0}^{x_1} y_1 \frac{x - x_0}{x_1 - x_0} dx$$
$$= y_0 \frac{h}{2} + y_1 \frac{h}{2} \quad (h = x_1 - x_0)$$
$$= h \frac{y_0 + y_1}{2}$$

$$\int_{x_0}^{x_1} E(x)dx = \frac{1}{2!} \int_{x_0}^{x_1} (x - x_0)(x - x_1) f'(c) dx$$

$$= \frac{f''(c)}{2} \int_{x_0}^{x_1} (x - x_0)(x - x_1) dx$$

$$= \frac{f''(c)}{2} \int_0^h u(u - h) du \quad (h = x_1 - x_0 \text{ and } u = x - x_0)$$

$$= -\frac{h^3}{12} f''(c)$$

Motivation

Definition (Trapezoid rule)

$$\int_{x_0}^{x_1} f(x) dx = \frac{h}{2} (y_0 + y_1) - \frac{h^3}{12} f''(c)$$

where $h = x_1 - x_0$ and $x_0 \le c \le x_1$.



Conclusions



Numerical differentiation Numerical Integration Conclusions

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Women's

- f: a function with a continuous fourth derivative on $[x_0, x_1]$.
- $y_0 = f(x_0)$ and $y_1 = f(x_1)$.





Motivation

- f: a function with a continuous fourth derivative on $[x_0, x_1]$.
- $y_0 = f(x_0)$ and $y_1 = f(x_1)$.
- $f(x) = y_0 \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} + y_1 \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} + y_2 \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_0-x_1)} + \frac{(x-x_0)(x-x_1)(x-x_2)}{3!} f'''(c).$
- P(x): $y_0 \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} + y_1 \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} + y_2 \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_0-x_1)}$;
- E(x): the error formula $\frac{(x-x_0)(x-x_1)(x-x_2)}{3!}f'''(c)$ where $x_0 < c < x_2$.





Motivation

• f: a function with a continuous fourth derivative on $[x_0, x_1]$.

- $y_0 = f(x_0)$ and $y_1 = f(x_1)$.
- $f(x) = y_0 \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} + y_1 \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} + y_2 \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_0-x_1)} + \frac{(x-x_0)(x-x_1)(x-x_2)}{3!} f'''(c).$
- P(x): $y_0 \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} + y_1 \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} + y_2 \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_0-x_1)}$;
- E(x): the error formula $\frac{(x-x_0)(x-x_1)(x-x_2)}{3!}f'''(c)$ where $x_0 \le c \le x_2$.
- $\int_{x_0}^{x_1} f(x) dx = \int_{x_0}^{x_1} P(x) dx + \int_{x_0}^{x_1} E(x) dx$.



$$\int_{x_0}^{x_2} P(x) dx = y_0 \int_{x_0}^{x_2} \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)} dx + y_1 \int_{x_0}^{x_2} \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)} dx + y_2 \int_{x_0}^{x_2} \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_0 - x_1)} dx$$





Simpson's rule

$$\int_{x_0}^{x_2} P(x)dx = y_0 \int_{x_0}^{x_2} \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)} dx + y_1 \int_{x_0}^{x_2} \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)} dx + y_2 \int_{x_0}^{x_2} \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_0 - x_1)} dx$$

$$= y_0 \frac{h}{3} + y_1 \frac{4h}{3} + y_2 \frac{h}{3} \quad (h = x_2 - x_1 = x_1 - x_0)$$

$$\int_{x_0}^{x_2} E(x) dx = -\frac{h^5}{90} f^{(iv)}(c)$$



Liangda Fang

Simpson's rule

Motivation

Definition (Simpson's rule)

$$\int_{x_0}^{x_2} f(x) dx = \frac{h}{3} (y_0 + 4y_1 + y_2) - \frac{h^5}{90} f^{(iv)}(c)$$

where $h = x_2 - x_1 = x_1 - x_0$ and $x_0 \le c \le x_2$.



Conclusions



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Liangda Fang

Apply the Trapezoid rule and Simpson's rule to approximate $\int_1^2 \ln x dx$, and find an upper bound for the error in your approximations.

•
$$\int_{1}^{2} \ln x dx = x \ln x \Big|_{1}^{2} - \int_{1}^{2} dx = 2 \ln 2 - 1 \ln 1 - 1 \approx 0.3863.$$



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Example (Trapezoid rule)

- $y_0 = 1$ and $y_1 = 2$.
- Approximation:

$$\int_{1}^{2} \ln x dx \approx \frac{h}{2} (y_0 + y_1) = \frac{1}{2} (\ln 1 + \ln 2) = \frac{\ln 2}{2} \approx 0.3466.$$

- Error: $-\frac{h^3}{12}f''(c) \le = \frac{1^3}{12c^2} \le \frac{1}{2} \approx 0.0834 \ (f''(x) = -\frac{1}{c^2}).$





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Example (Simpson's rule)

- $y_0 = 1$, $y_1 = \frac{3}{2}$ and $y_2 = 2$.
- Approximation:

$$\int_{1}^{\frac{1}{2}} \ln x dx \approx \frac{h}{3} (y_0 + 4y_1 + y_2) = \frac{0.5}{3} (\ln 1 + 4 \ln \frac{3}{2} \ln 2) = \frac{\ln 2}{2} \approx 0.3858.$$

- Error: $-\frac{h^5}{90}f^{(iv)}(c) \leq \frac{6(0.5)^5}{90c^4} \leq \frac{6(0.5)^5}{90} = \frac{1}{490} \approx 0.0021.$
- $\int_{1}^{2} \ln x dx = 0.3858 \pm 0.0021.$





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Composite Newton-Cotes formulas (Composite Trapezoid Rule)

- f: a function with a continuous second derivative on [a, b].
- evenly spaced grid: $a = x_0 < x_1 < \cdots < x_{m-1} < x_m = b$.
- $y_0 = f(x_0) = f(a)$ and $y_m = f(x_m) = f(b)$.

Motivation



Conclusions



Liangda Fang 35/83

Composite Trapezoid Rule

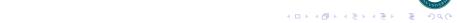
Motivation

Apply the **Trapezoid Rule** separately on each subinterval $[x_i, x_{i+1}]$:

$$\int_{x_i}^{x_{i+1}} f(x) dx = \frac{h}{2} (f(x_i) + f(x_{i+1})) - \frac{h^3}{12} f'(c_i),$$
where $h = x_{i+1} - x_i$ and $x_i \le c_i \le x_{i+1}$.



Conclusions



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Apply the **Trapezoid Rule** separately on each subinterval $[x_i, x_{i+1}]$:

$$\int_{x_i}^{x_{i+1}} f(x) dx = \frac{h}{2} (f(x_i) + f(x_{i+1})) - \frac{h^3}{12} f'(c_i),$$
 where $h = x_{i+1} - x_i$ and $x_i \le c_i \le x_{i+1}$.

Total up over all subintervals:

$$\int_{a}^{b} f(x) dx = \sum_{i=0}^{m-1} \int_{x_{i}}^{x_{i+1}} f(x) dx$$
$$= \frac{h}{2} [f(a) + f(b) + 2 \sum_{i=1}^{m-1} f(x_{i})] - \sum_{i=0}^{m-1} \frac{h^{3}}{12} f'(c_{i}).$$





Liangda Fang 36/83

Composite Trapezoid Rule

Motivation

Apply the **Trapezoid Rule** separately on each subinterval $[x_i, x_{i+1}]$:

$$\int_{x_i}^{x_{i+1}} f(x) dx = \frac{h}{2} (f(x_i) + f(x_{i+1})) - \frac{h^3}{12} f'(c_i),$$
 where $h = x_{i+1} - x_i$ and $x_i \le c_i \le x_{i+1}$.

Total up over all subintervals:

$$\int_{a}^{b} f(x) dx = \sum_{i=0}^{m-1} \int_{x_{i}}^{x_{i+1}} f(x) dx$$
$$= \frac{h}{2} [f(a) + f(b) + 2 \sum_{i=1}^{m-1} f(x_{i})] - \sum_{i=0}^{m-1} \frac{h^{3}}{12} f'(c_{i}).$$

According to **Generalized Intermediate Value Theorem**, the error term can be written:

$$\sum_{i=0}^{m-1} \tfrac{h^3}{12} f'(c_i) = \tfrac{h^3}{12} m f'(c), \text{ where } a < c < b.$$



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Composite Trapezoid Rule

Motivation

Definition (Composite Trapezoid Rule)

$$\int_{a}^{b} f(x) dx = \frac{h}{2} (y_0 + y_m + 2 \sum_{i=1}^{m-1} y_i) - \frac{(b-a)h^2}{12} f''(c)$$

where h = (b - a)/m and $a \le c \le b$.



Liangda Fang 37/83

Composite Newton-Cotes formulas (Composite Simpson's Rule)

- f: a function with a continuous fourth derivative on [a, b].
- evenly spaced grid: $a = x_0 < x_1 < \cdots < x_{2m-1} < x_{2m} = b$.
- $y_0 = f(x_0) = f(a)$ and $y_{2m} = f(x_{2m}) = f(b)$.



Conclusions



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Composite Simpson's Rule

Apply the **Simpson's Rule** separately on each subinterval $[x_{2i}, x_{2i+2}]$:

$$\int_{x_{2i}}^{x_{2i+2}} f(x) dx = \frac{h}{3} [f(x_{2i}) + 4f(x_{2i+1}) + f(x_{2i+2})] - \frac{h^5}{90} f^{(iv)}(c_i),$$
where $h = x_{i+1} - x_i$ and $x_{2i} \le c_i \le x_{2i+2}$.



Liangda Fang 39/83

Composite Simpson's Rule

Apply the **Simpson's Rule** separately on each subinterval $[x_{2i}, x_{2i+2}]$:

$$\int_{x_{2i}}^{x_{2i+2}} f(x) dx = \frac{h}{3} [f(x_{2i}) + 4f(x_{2i+1}) + f(x_{2i+2})] - \frac{h^5}{90} f^{(iv)}(c_i),$$
where $h = x_{i+1} - x_i$ and $x_{2i} \le c_i \le x_{2i+2}$.

Total up over all subintervals:

$$\int_{a}^{b} f(x) dx = \sum_{i=0}^{m-1} \int_{x_{2i}}^{x_{2i+2}} f(x) dx$$

$$= \frac{h}{3} [f(a) + f(b) + 4 \sum_{i=1}^{m} f(x_{2i-1}) + 2 \sum_{i=1}^{m-1} f(x_{2i})] - \sum_{i=0}^{m-1} \frac{h^{5}}{90} f^{(iv)}(c_{i})$$

According to **Generalized Intermediate Value Theorem**, the error term can be written:

$$\sum_{i=0}^{m-1} \frac{h^5}{90} f^{(iv)}(c_i) = \frac{h^5}{90} m f^{(iv)}(c), \text{ where } a < c < b.$$



Liangda Fang 39/83

Composite Simpson's Rule

Motivation

Definition (Composite Simpson's Rule)

$$\int_{a}^{b} f(x) dx = \frac{h}{3} [y_0 + y_{2m} + 4 \sum_{i=1}^{m} y_{2i-1} + 2 \sum_{i=1}^{m-1} y_{2i}] - \frac{(b-a)h^4}{180} f^{(iv)}(c)$$

where h = (b - a)/2m and $a \le c \le b$.



Conclusions



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Composite Newton-Cotes formulas (Example)

Example

Motivation

Carry out **four-panel** approximations of $\int_1^2 \ln x dx$ using the composite Trapezoid Rule and composite Simpson's Rule.

•
$$\int_{1}^{2} \ln x dx = x \ln x \Big|_{1}^{2} - \int_{1}^{2} dx = 2 \ln 2 - 1 \ln 1 - 1 \approx 0.3863.$$



Conclusions



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Example (composite Trapezoid rule)

- h = 1/4, $y_0 = \ln 1$, $y_1 = \ln \frac{5}{4}$, $y_2 = \ln \frac{3}{2}$, $y_3 = \ln \frac{7}{4}$ and $y_4 = \ln 2$.
- Approximation: $\int_{1}^{2} \ln x dx \approx \frac{h}{2} (y_0 + y_4 + 2 \sum_{i=1}^{3} y_i) = \frac{1}{8} [\ln 1 + \ln 2 + 2(\ln \frac{5}{4} + \ln \frac{3}{2} + \ln \frac{7}{4})] \approx 0.3837.$
- $\bullet \ \, \mathsf{Error} \colon \, -\tfrac{(b-a)h^2}{12} f''(c) \le \tfrac{1/16}{12c^2} \le \tfrac{1}{(16)(12)(1^2)} \approx 0.0052. \, \left(f''(x) = -\tfrac{1}{x^2}\right)$
- $\int_{1}^{2} \ln x dx = 0.3837 \pm 0.0052.$



Liangda Fang 42/83

Example (composite Simpson's rule)

- h = 1/8, $y_0 = \ln 1$, $y_1 = \ln \frac{9}{8}$, $y_2 = \ln \frac{5}{4}$, $y_3 = \ln \frac{11}{8}$, $y_4 = \ln \frac{6}{4}$, $y_5 = \ln \frac{13}{8}$, $y_6 = \ln \frac{7}{4}$, $y_7 = \ln \frac{15}{8}$ and $y_8 = \ln 2$.
- Approximation:

$$\int_{1}^{2} \ln x dx \approx \frac{1/8}{3} [y_0 + y_8 + 4 \sum_{i=1}^{8} y_{2i-1} + 2 \sum_{i=1}^{7} y_{2i}] = \frac{1}{24} [\ln 1 + \ln 2 + 4(\ln \frac{9}{8} + \ln \frac{11}{8} + \ln \frac{13}{8} + \ln \frac{15}{8}) + 2(\ln \frac{5}{4} + \ln \frac{6}{4} + \ln \frac{7}{4})] \approx 0.386292.$$

- $\bullet \ \, \mathsf{Error:} \,\, -\tfrac{(b-a)h^4}{180} \mathit{f}^{(iv)}(c) \leq \tfrac{6}{8^4 \cdot 180 \cdot 1^4} \approx 0.000008. \,\, \left(\mathit{f}^{(iv)}(x) = -\tfrac{6}{x^4}\right)$
- $\int_{1}^{2} \ln x dx = 0.386292 \pm 0.000008.$



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Open Newton-Cotes Methods (Midpoint Rule)

- f: a function with a continuous second derivative on $[x_0, x_1]$.
- $h = x_1 x_0$ and $w = x_0 + \frac{h}{2}$.





Liangda Fang

Open Newton-Cotes Methods (Midpoint Rule)

- f: a function with a continuous second derivative on $[x_0, x_1]$.
- $h = x_1 x_0$ and $w = x_0 + \frac{h}{2}$.

Motivation

The degree 1 Taylor expansion of f(x) about the midpoint w:

$$f(x) = f(w) + (x - w)f(w) + \frac{(x-w)^2}{2}f'(c_x)$$
, where $x_0 < c_x < x_1$.



Conclusions



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Open Newton-Cotes Methods (Midpoint Rule)

- f: a function with a continuous second derivative on $[x_0, x_1]$.
- $h = x_1 x_0$ and $w = x_0 + \frac{h}{2}$.

The degree 1 Taylor expansion of f(x) about the midpoint w:

$$f(x) = f(w) + (x - w)f(w) + \frac{(x-w)^2}{2}f'(c_x)$$
, where $x_0 < c_x < x_1$.

Integrate both sides:

$$\begin{split} \int_{x_0}^{x_1} f(x) \, dx &= h f(w) + f'(w) \int_{x_0}^{x_1} (x-w) \, dx + \frac{1}{2} \int_{x_0}^{x_1} f''(c_x) (x-w)^2 \, dx \\ &= h f(w) + 0 + \frac{f'(c)}{2} \int_{x_0}^{x_1} (x-w)^2 \, dx \text{ (Mean Value Theorem)} \\ &= h f(w) + \frac{h^3}{24} f''(c) \text{ , where } x_0 < c < x_1 \end{split}$$



Liangda Fang 45/8

Midpoint Rule

Definition (Midpoint Rule)

$$\int_{x_0}^{x_1} f(x) dx = hf(w) + \frac{h^3}{24} f'(c)$$

where $h = x_1 - x_0$, $w = x_0 + \frac{h}{2}$ and $x_0 < c < x_1$.

- does not use values from the endpoints;
- cut the number of function evaluations needed;
- the error term is half the size of the Trapezoid Rule error term.



Liangda Fang 46/83

- f: a function with a continuous second derivative on [a, b].
- evenly spaced grid: $a = x_0 < x_1 < \cdots < x_{m-1} < x_m = b$.





Liangda Fang

Open Newton-Cotes Methods (Composite Midpoint Rule)

- f: a function with a continuous second derivative on [a, b].
- evenly spaced grid: $a = x_0 < x_1 < \cdots < x_{m-1} < x_m = b$.

Definition (Composite Midpoint Rule)

$$\int_{a}^{b} f(x) dx = h \sum_{i=1}^{m} f(w_i) + \frac{(b-a)h^2}{24} f''(c)$$

where h = (b-a)/m, $w_i = \frac{(x_{i-1}+x_i)}{2}$ for $1 \le i \le m$ and a < c < b.





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Motivation

Approximate $\int_0^1 \frac{\sin x}{x} dx$ by using the Composite Midpoint Rule with m=10 panels.

- The correct answer to eight places of $\int_0^1 \frac{\sin x}{x} dx$ is 0.94608307.
- $h = 0.1, \{w_1, w_2, \dots, w_{10}\} = \{0.05, 0.15, \dots, 0.95\}.$
- Approximation: $\int_0^1 \frac{\sin x}{x} dx \approx 0.1 \sum_{i=1}^{10} f(w_i) \approx 0.94620858.$



Conclusions



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Motivation Numerical differentiation Numerical Integration Conclusions

Outline

- Motivation
- 2 Numerical differentiation
 - Finite difference formulas
 - Extrapolation
- 3 Numerical Integration
 - Trapezoid rule
 - Simpson's rule
 - Composite Newton-Cotes formulas
 - Open Newton-Cotes Methods
 - Romberg Integration
 - Adaptive Quadrature
 - Gaussian Quadrature
- 4 Conclusions





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Romberg Integration

A method to apply extrapolation to the composite Trapezoid Rule.

Definition (Extrapolation)

$$M = R(h) + K_n h^n + K_{n+1} h^{n+1} + \cdots,$$

where K_n, K_{n+1}, \cdots are constants.

We say R(h) is a nth-order approximation of a given quantity M.





Romberg Integration

A method to apply extrapolation to the composite Trapezoid Rule.

Definition (Extrapolation)

$$M = R(h) + K_n h^n + K_{n+1} h^{n+1} + \cdots,$$

where K_n, K_{n+1}, \cdots are constants.

We say R(h) is a nth-order approximation of a given quantity M.

Definition (Composite Trapezoid Rule)

Let f be an infinitely differentiable function.

$$\int_{a}^{b} f(x) dx = \frac{h}{2} (y_0 + y_m + 2 \sum_{i=1}^{m-1} y_i) + c_2 h^2 + c_4 h^4 + c_6 h^6 + \cdots$$

where h=(b-a)/m and c_i is a constant that depends only on higher derivatives of f at a and b.

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Romberg Integration

$$\int_{a}^{b} f(x) dx = \frac{h}{2} (y_0 + y_m + 2 \sum_{i=1}^{m-1} y_i) + c_2 h^2 + c_4 h^4 + c_6 h^6 + \cdots$$

Example (Single extrapolation)

- $h_1 = b a, h_2 = \frac{h_1}{2}$
- Approximation $R_{11}(h_1)$: $\frac{h_1}{2}[f(a)+f(b)]$
- \bullet Approximation $R_{21}(h_2)$: $\frac{h_2}{2}[\mathit{f}(a)+\mathit{f}(b)+2\mathit{f}(\frac{a+b}{2})]$
- $\int_a^b f(x) dx = R_{21}(h_2) + \frac{c_2 h_2^2}{2} + O(h_2^4)$
- Combining the above equations gives the following: $\int_a^b f(x) dx = \frac{4}{3} R_{21}(h_2) \frac{1}{3} R_{11}(h_1) + \frac{O(h_1^4)}{2}$



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Motivation

Table: The Romberg Table

| Error | $O(h_1^2)$ | $O(h_1^4)$ | $O(h_1^6)$ | | $O(h_1^{2j})$ |
|-------|------------|------------|------------|---|---------------|
| h_1 | R_{11} | | | | |
| h_2 | R_{21} | R_{22} | | | |
| h_3 | R_{31} | R_{32} | R_{33} | | |
| : | : | : | : | ٠ | |
| h_j | R_{j1} | R_{j2} | R_{j3} | | R_{jj} |

- The first column R_{j1} is the approximations using the composite Trapezoid Rule at step size h_{j} . (2th-order approximations)
- The second column R_{j2} is the extrapolations of the first column R_{j1} . (4th-order approximations)
- Similarly, the k-th column R_{jk} is the extrapolations of the (k-1)-th column column $R_{j,k-1}$. (2kth-order approximations)
- The best approximation for $\int_a^b f(x) dx$ is R_{jj} . (2*j*th-order approximations)

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Romberg Integration (calculate composite Trapezoid Rule incrementally)

$$\int_a^b f(x)dx = \frac{h}{2}(y_0 + y_m + 2\sum_{i=1}^{m-1} y_i) + c_2h^2 + c_4h^4 + c_6h^6 + \cdots$$

- $h_j = \frac{(b-a)}{2^{j-1}}$ for j = 2, 3, ...
- Approximation using h_j : $R_{j1} = \frac{h_j}{2}(f(a) + f(b) + 2\sum_{i=1}^{2^{j-1}-1} y_i)$.





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$$\int_{a}^{b} f(x) dx = \frac{h}{2} (y_0 + y_m + 2 \sum_{i=1}^{m-1} y_i) + c_2 h^2 + c_4 h^4 + c_6 h^6 + \cdots$$

• $h_j = \frac{(b-a)}{2^{j-1}}$ for j = 2, 3, ...

Motivation

 $\begin{array}{ll} \bullet \ \ \text{Approximation using} \ h_j: \ R_{j1} = \frac{h_j}{2} (f(a) + f(b) + 2 \sum_{i=1}^{2^{j-1}-1} y_i). \\ \text{e.g.} \\ R_{11} & = \frac{h_1}{2} [f(a) + f(b)], \\ R_{21} & = \frac{h_2}{2} [f(a) + f(b) + 2 f(\frac{a+b}{2})] \\ & = \frac{1}{2} R_{11} + h_2 f(\frac{a+b}{2}) \end{array}$



Conclusions



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Romberg Integration (calculate composite Trapezoid Rule incrementally)

$$\int_a^b f(x) dx = \frac{h}{2} (y_0 + y_m + 2 \sum_{i=1}^{m-1} y_i) + c_2 h^2 + c_4 h^4 + c_6 h^6 + \cdots$$

- $h_i = \frac{(b-a)}{2^{i-1}}$ for j = 2, 3, ...
- Approximation using h_j : $R_{j1} = \frac{h_j}{2}(f(a) + f(b) + 2\sum_{i=1}^{2^{j-1}-1} y_i)$. e.g. $R_{11} = \frac{h_1}{2}[f(a) + f(b)],$ $R_{21} = \frac{h_2}{2}[f(a) + f(b) + 2f(\frac{a+b}{2})]$ $= \frac{1}{2}R_{11} + h_2 f(\frac{a+b}{2})$

It can be written by induction as follows:

$$R_{j1} = \frac{1}{2}R_{j-1,1} + h_j \sum_{i=1}^{2^{j-2}} f(a + (2j-1)h_j)$$

, for each $i=2,3,\ldots$



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Romberg Integration (extrapolation)

| Error | $O(h_1^2)$ | $O(h_1^4)$ | $O(h_1^6)$ | | $O(h_1^{2j})$ | |
|--|------------|------------|------------|-------|---------------|--|
| h_1 | R_{11} | | | | | |
| h_2 | R_{21} | R_{22} | | | | |
| h_3 | R_{31} | R_{32} | R_{33} | | | |
| : | : | : | : | ٠ | | |
| h_j | R_{j1} | R_{j2} | R_{j3} | • • • | R_{jj} | |
| $R_{22} = \frac{2^{2}R_{21} - R_{11}}{3}, R_{33} = \frac{4^{2}R_{32} - R_{22}}{4^{2} - 1}, \cdots$ $R_{32} = \frac{2^{2}R_{31} - R_{21}}{3}, R_{43} = \frac{4^{2}R_{42} - R_{32}}{4^{2} - 1}, \cdots$ $R_{42} = \frac{2^{2}R_{41} - R_{31}}{3}, R_{53} = \frac{4^{2}R_{52} - R_{42}}{4^{2} - 1}, \cdots$ | | | | | | |





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Romberg Integration (extrapolation)

| Error | $O(h_1^2)$ | $O(h_1^4)$ | $O(h_1^6)$ | | $O(h_1^{2j})$ | |
|---|------------|------------|------------|---|---------------|--|
| h_1 | R_{11} | | | | | |
| h_2 | R_{21} | R_{22} | | | | |
| h_3 | R_{31} | R_{32} | R_{33} | | | |
| : | : | ÷ | ÷ | ٠ | | |
| h_j | R_{j1} | R_{j2} | R_{j3} | | R_{jj} | |
| $R_{22} = \frac{2^2 R_{21} - R_{11}}{3}, R_{33} = \frac{4^2 R_{32} - R_{22}}{4^2 - 1}, \cdots$ $R_{32} = \frac{2^2 R_{31} - R_{21}}{3}, R_{43} = \frac{4^2 R_{42} - R_{32}}{4^2 - 1}, \cdots$ | | | | | | |
| $R_{42} = \frac{2^2 R_{41} - R_{31}}{3}, R_{53} = \frac{4^2 R_{52} - R_{42}}{4^2 - 1}, \dots$ | | | | | | |

ullet Extrapolations of the approximations $R_{j,k-1}$ and $R_{j-1,k-1}$:

$$R_{jk} = \frac{1}{4^{k-1} - 1} (4^{k-1} R_{j,k-1} - R_{j-1,k-1})$$

, for each $j=2,3,\ldots$ and $k=j,j+1,\ldots$



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Romberg Integration

Example

Apply Romberg Integration to approximate $\int_{1}^{2} \ln x dx$.

• $\int_{1}^{2} \ln x dx = x \ln x \Big|_{1}^{2} - \int_{1}^{2} dx = 2 \ln 2 - 1 \ln 1 - 1 \approx 0.38629436.$

| j | h_j | R_{j1} | R_{j2} | R_{j3} | R_{jj} |
|---|-------|------------------|------------------|------------------|------------------|
| 1 | 1 | 0.34657359027997 | | | |
| 2 | 0.5 | 0.37601934919407 | 0.38583460216543 | | |
| 3 | 0.25 | 0.38369950940944 | 0.38625956281457 | 0.38628789352451 | |
| 4 | 0.125 | 0.38564390995210 | 0.38629204346631 | 0.38629420884310 | 0.38629430908625 |

- R_{11} : Approximation using the Trapezoid Rule;
- R₂₂: Approximation using the Simpson's Rule;
- ullet R_{31} : four-panel approximation using the composite Trapezoid Rule;
- ullet R_{42} : four-panel approximation using the composite Simpson's Rule;



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Motivation Numerical differentiation Numerical Integration Conclusions

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- Motivation
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Adaptive Quadrature (Adaptive Trapezoid Rule Quadrature)

Adaptive Quadrature:

- Variable step size.
- 2 The integration error is estimated by the integration approximations rather than by the error formulas directly.
 - f: a function with a continuous second derivative on [a, b].
 - h = b a.
 - c: the midpoint of [a,b].
 - $S_{[a,b]}$: the approximation for $\int_a^b f(x) dx$ using Trapezoid Rule on the interval [a,b].
- $S_{[a,c]}$ (resp. $S_{[c,b]}$): the approximation for $\int_a^b f(x) dx$ using Trapezoid Rule on the interval [a,c] (resp. [c,b]).

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Apply Trapezoid Rule on interval [a,b] and half-intervals [a,c] and [c,b]:

$$\int_{a}^{b} f(x) dx = S_{[a,b]} - h^{3} \frac{f''(c_{0})}{12}, \text{ where } a < c_{0} < b$$



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Apply Trapezoid Rule on interval [a,b] and half-intervals [a,c] and [c,b]:

$$\begin{split} \int_a^b f(x) \, dx &= S_{[a,b]} - h^3 \frac{f''(c_0)}{12}, \text{ where } a < c_0 < b \\ \int_a^b f(x) \, dx &= S_{[a,c]} - \frac{h^3}{8} \frac{f''(c_1)}{12} + S_{[c,b]} - \frac{h^3}{8} \frac{f''(c_2)}{12} \\ &= S_{[a,c]} + S_{[c,b]} - \frac{h^3}{4} \frac{f''(c_3)}{12} \\ &\text{where } a < c_1 < c, c < c_2 < b \text{ and } a < c_3 < b \end{split}$$





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Apply Trapezoid Rule on interval [a,b] and half-intervals [a,c] and [c,b]:

$$\begin{split} \int_a^b f(x) \, dx &= S_{[a,b]} - h^3 \frac{f''(c_0)}{12}, \text{ where } a < c_0 < b \\ \int_a^b f(x) \, dx &= S_{[a,c]} - \frac{h^3}{8} \frac{f''(c_1)}{12} + S_{[c,b]} - \frac{h^3}{8} \frac{f''(c_2)}{12} \\ &= S_{[a,c]} + S_{[c,b]} - \frac{h^3}{4} \frac{f''(c_3)}{12} \\ &\text{where } a < c_1 < c, c < c_2 < b \text{ and } a < c_3 < b \end{split}$$

Subtracting the above two equations gives the following:

$$S_{[a,b]} - (S_{[a,c]} + S_{[c,b]}) = -\frac{h^3}{4} \frac{f''(c_3)}{12} + h^3 \frac{f''(c_0)}{12}$$

$$\approx \frac{3}{4} h^3 \frac{f''(c_3)}{12}, \text{ where } f''(c_3) \approx f''(c_0)$$

It is approximately three times the size of the integration error of the approximation $S_{[a,c]} + S_{[c,b]}$.



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Algorithm 1: Adaptive Trapezoid Rule Quadrature

// To approximate $\int_a^b f(x) dx$ within tolerance TOL:

- 1 $c = \frac{a+b}{2}$
- 2 $S_{[a,b]} = (b-a) \frac{f(a)+f(b)}{2}$
- 3 if $|S_{[a,b]} (S_{[a,c]} + S_{[c,b]}| < 3 \cdot TOL \cdot (\frac{b-a}{b_{orin} a_{orin}})$ then
- 4 \bigsqcup accept $S_{[a,c]}+S_{[c,b]}$ as approximation over [a,b]
- 5 else
- ${f 6}$ repeat above recursively for [a,c] and [c,b]
 - Every time breaking intervals in half, the required error tolerance for the subinterval goes down by a factor of 2.
 - Line 3 is to check whether $S_{[a,c]} + S_{[c,b]}$ approximates the unknown exact integral within $TOL \cdot (\frac{b-a}{b_{orio}-a_{orio}})$.



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Adaptive Quadrature (Adaptive Simpson's Rule Quadrature)

- f: a function with a continuous fourth derivative on [a, b].
- h = b a.
- c: the midpoint of [a, b].
- $S_{[a,b]}$ (resp. $S_{[a,c]}$, $S_{[c,b]}$): the approximation for $\int_a^b f(x) dx$ using Simpson's Rule on the interval [a, b] (resp. [a, c], [c, b]).





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Adaptive Simpson's Rule Quadrature

Apply Simpson's Rule on interval [a, b] and its two halves:

$$\int_{a}^{b} f(x) dx = S_{[a,b]} - h^{5} \frac{f^{(iv)}(c_{0})}{90}, \text{ where } a < c_{0} < b$$





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Adaptive Simpson's Rule Quadrature

Apply Simpson's Rule on interval [a, b] and its two halves:

$$\begin{split} \int_a^b f(x) dx &= S_{[a,b]} - h^5 \frac{f^{(iv)}(c_0)}{90}, \text{ where } a < c_0 < b \\ \int_a^b f(x) dx &= S_{[a,c]} - \frac{h^5}{32} \frac{f^{(iv)}(c_1)}{90} + S_{[c,b]} - \frac{h^5}{32} \frac{f^{(iv)}(c_2)}{90} \\ &= S_{[a,c]} + S_{[c,b]} - \frac{h^5}{16} \frac{f^{(iv)}(c_3)}{90} \\ &\text{where } a < c_1 < c, c < c_2 < b \text{ and } a < c_3 < b \end{split}$$



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Adaptive Simpson's Rule Quadrature

Apply Simpson's Rule on interval [a, b] and its two halves:

$$\begin{split} \int_a^b f(x) \, dx &= S_{[a,b]} - h^5 \frac{f^{(iv)}(c_0)}{90}, \text{ where } a < c_0 < b \\ \int_a^b f(x) \, dx &= S_{[a,c]} - \frac{h^5}{32} \frac{f^{(iv)}(c_1)}{90} + S_{[c,b]} - \frac{h^5}{32} \frac{f^{(iv)}(c_2)}{90} \\ &= S_{[a,c]} + S_{[c,b]} - \frac{h^5}{16} \frac{f^{(iv)}(c_3)}{90} \\ &\text{where } a < c_1 < c, c < c_2 < b \text{ and } a < c_3 < b \end{split}$$

Subtracting the above two equations gives the following:

$$\begin{split} S_{[a,b]} - (S_{[a,c]} + S_{[c,b]}) &= h^5 \frac{f^{(iv)}(c_0)}{90} - \frac{h^5}{16} \frac{f^{(iv)}(c_3)}{90} \\ &\approx \frac{15}{16} h^5 \frac{f^{(iv)}(c_3)}{90}, \text{ where } f^{(iv)}(c_3) \approx f^{(iv)}(c_0) \end{split}$$

It is **approximately 15 times** the size of the integration error of the approximation $S_{[a,c]} + S_{[c,b]}$.

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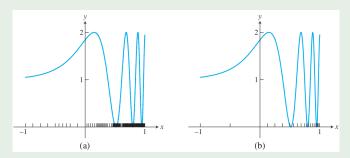
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Adaptive Quadrature

Numerical differentiation

Example

Use Adaptive Quadrature to approximate the integral $\int_{-1}^{1} (1 + \sin e^{3x}) dx$.



- Tolerance TOL = 0.005.
- (a) Adaptive Trapezoid Rule requires 140 subintervals.
- (b) Adaptive Simpson's Rule requires 20 subintervals.

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Motivation Numerical differentiation Numerical Integration Conclusions

Outline

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Gaussian Quadrature (Orthogonality)

Definition (Orthogonality)

The set of nonzero functions $\{p_0, p_1, \cdots, p_n\}$ on the interval [a, b] is **orthogonal** on [a, b] if

$$\int_{a}^{b} p_{j}(x) p_{k}(x) dx = \begin{cases} 0 & , j \neq k \\ \neq 0 & , j = k \end{cases}$$

Theorem

If $\{p_0, p_1, \dots, p_n\}$ is an orthogonal set of polynomials on the interval [a, b], where deg $p_i = i$, then:

- $\{p_0, p_1, \dots, p_n\}$ is a basis for the vector space of degree at most n polynomials on [a, b].
- p_i has i distinct roots in the interval (a, b).



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Orthogonality

Example

Find a set of three orthogonal polynomials on the interval [-1,1].

•
$$p_0(x) = 1, p_1(x) = x, p_2(x) = x^2 + c$$



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Orthogonality

Example

Find a set of three orthogonal polynomials on the interval [-1,1].

- $p_0(x) = 1, p_1(x) = x, p_2(x) = x^2 + c$
- p_0 and p_1 are orthogonal:

$$\int_{-1}^{1} p_0(x)p_1(x)dx = \int_{-1}^{1} 1 \cdot xdx = \frac{1}{2}x^2|_{-1}^{1} = 0$$

• p_1 and p_2 are orthogonal:

$$\int_{-1}^{1} p_1(x)p_2(x)dx = \int_{-1}^{1} x \cdot (x^2 + c)dx = (\frac{1}{4}x^4 + \frac{c}{2}x^2)|_{-1}^{1} = 0$$



Conclusions

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Orthogonality

Example

Find a set of three orthogonal polynomials on the interval [-1,1].

- $p_0(x) = 1, p_1(x) = x, p_2(x) = x^2 + c$
- p_0 and p_1 are orthogonal:

$$\int_{-1}^{1} p_0(x)p_1(x)dx = \int_{-1}^{1} 1 \cdot xdx = \frac{1}{2}x^2|_{-1}^{1} = 0$$

ullet p_1 and p_2 are orthogonal:

$$\int_{-1}^{1} p_1(x)p_2(x)dx = \int_{-1}^{1} x \cdot (x^2 + c)dx = (\frac{1}{4}x^4 + \frac{c}{2}x^2)|_{-1}^{1} = 0$$

• when $c = \frac{-1}{3}$, p_0 and p_2 are orthogonal:

$$\int_{-1}^{1} p_0(x) p_2(x) dx = \int_{-1}^{1} 1 \cdot (x^2 + c) dx = \frac{2}{3} + 2c = 0$$

• the set $\{1, x, x^2 - 1/3\}$ is an orthogonal set on [-1, 1].

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Gaussian Quadrature (Legendre polynomials)

The set of Legendre polynomials

$$p_i(x) = \frac{1}{2^i i!} \frac{d^i}{dx^i} [(x^2 - 1)^i]$$

for $0 \le i \le n$ is orthogonal on [-1, 1].





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Gaussian Quadrature (Legendre polynomials)

The set of Legendre polynomials

$$p_i(x) = \frac{1}{2^i i!} \frac{d^i}{dx^i} [(x^2 - 1)^i]$$

for $0 \le i \le n$ is orthogonal on [-1, 1].

- $\int_{-1}^{1} p_i(x) p_i(x) dx = \frac{2}{2i+1} \neq 0$ for $0 \leq i \leq n$.
- $\int_{-1}^{1} p_i(x) p_j(x) dx = 0$ for $0 \le i \ne j \le n$.





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Proof of the second item.

We show that if i < j, then $\int_{-1}^{1} [(x^2 - 1)^i]^{(i)} [(x^2 - 1)^j]^{(j)} dx = 0$.

• Integrate by parts with $u = [(x^2 - 1)^i]^{(i)}$ and $dv = [(x^2 - 1)^j]^{(j)} dx$.

$$uv - \int_{-1}^{1} v du = [(x^{2} - 1)^{i}]^{(i)}[(x^{2} - 1)^{j}]^{(j-1)}|_{-1}^{1}$$
$$- \int_{-1}^{1} [(x^{2} - 1)^{i}]^{(i+1)}[(x^{2} - 1)^{j}]^{(j-1)} dx$$
$$= - \int_{-1}^{1} [(x^{2} - 1)^{i}]^{(i+1)}[(x^{2} - 1)^{j}]^{(j-1)} dx$$

since $[(x^2-1)^j]^{(j-1)}$ is divisible by x^2-1 .



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Legendre polynomials

Proof of the second item.

We show that if i < j, then $\int_{-1}^{1} [(x^2 - 1)^i]^{(i)} [(x^2 - 1)^j]^{(j)} dx = 0$.

• Integrate by parts with $u = [(x^2 - 1)^i]^{(i)}$ and $dv = [(x^2 - 1)^j]^{(j)} dx$.

$$uv - \int_{-1}^{1} v du = [(x^{2} - 1)^{i}]^{(i)}[(x^{2} - 1)^{j}]^{(j-1)}|_{-1}^{1}$$
$$- \int_{-1}^{1} [(x^{2} - 1)^{i}]^{(i+1)}[(x^{2} - 1)^{j}]^{(j-1)} dx$$
$$= - \int_{-1}^{1} [(x^{2} - 1)^{i}]^{(i+1)}[(x^{2} - 1)^{j}]^{(j-1)} dx$$

since $[(x^2-1)^j]^{(j-1)}$ is divisible by x^2-1 .

ullet After i+1 repeated integration by parts, we get that:

$$(-1)^{i+1} \int_{-1}^{1} \left[(x^2 - 1)^i \right]^{(2i+1)} \left[(x^2 - 1)^j \right]^{(j-i-1)} dx = 0$$

because the (2i+1)st derivative of $(x^2-1)^i$ is zero.



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Gaussian Quadrature of a function is simply a linear combination of function evaluations at the Legendre roots.

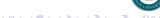


Liangda Fang 68/83

Gaussian Quadrature of a function is simply a linear combination of function evaluations at the Legendre roots.

- x_1, \dots, x_n : n distinct roots in [-1, 1] of the nth Legendre polynomial $p_n(x)$.
- Q(x): the Lagrange interpolating polynomial for the integrand f(x) at the nodes x_1, \dots, x_n .





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Using the Lagrange formulation, we can write:

$$Q(x) = \sum_{i=1}^{n} L_i(x) f(x_i), \text{ where } L_i(x) = \frac{(x-x_1)\cdots\overline{(x-x_i)}(x-x_n)}{(x_i-x_1)\cdots\overline{(x_i-x_i)}(x_i-x_n)}$$

Integrating both sides yields the following approximation for the integral:

$$\int_{-1}^{1} f(x) dx \approx \sum_{i=1}^{n} c_{i} f(x_{i}), \text{ where } c_{i} = \int_{-1}^{1} L_{i}(x) dx, i = 1, \cdots, n$$

coefficients c_i roots x_i $-\sqrt{1/3} = -0.57735026918963$ $\sqrt{1/3} = 0.57735026918963$ = 1.0000000000000000 $-\sqrt{3/5} = -0.77459666924148$ 5/9 = 0.55555555555558/9 = 0.888888888888880.000000000000000 $\sqrt{3/5} = 0.77459666924148$ 5/9 = 0.5555555555555 $-\sqrt{\frac{15+2\sqrt{30}}{35}} = -0.86113631159405$ $\frac{90-5\sqrt{30}}{100} = 0.34785484513745$ $-\sqrt{\frac{15-2\sqrt{30}}{35}} = -0.33998104358486$ $\frac{90+5\sqrt{30}}{180} = 0.65214515486255$ $\sqrt{\frac{15-2\sqrt{30}}{25}} = 0.33998104358486$ $\frac{90+5\sqrt{30}}{180} = 0.65214515486255$ $\frac{15+2\sqrt{30}}{25} = 0.86113631159405$ $\frac{90-5\sqrt{30}}{180} = 0.34785484513745$





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Example

Approximate $\int_{-1}^{1} e^{-\frac{x^2}{2}} dx$, using Gaussian Quadrature.

- $f(x) = e^{-\frac{x^2}{2}}$.
- The correct answer to 14 digits is 1.71124878378430.
- The n=2 approximation:

$$\int_{-1}^{1} e^{-\frac{x^2}{2}} dx \approx c_1 f(x_1) + c_2 f(x_2) + c_3 f(x_3)$$

$$= 1 \cdot f(-\sqrt{1/3}) + 1 \cdot f(\sqrt{1/3})$$

$$\approx 1.69296344978123$$



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Example

Approximate $\int_{-1}^{1} e^{-\frac{x^2}{2}} dx$, using Gaussian Quadrature.

• The n=3 approximation:

$$\int_{-1}^{1} e^{-\frac{x^2}{2}} dx \approx c_1 f(x_1) + c_2 f(x_2) + c_3 f(x_3)$$

$$= \frac{5}{9} \cdot f(-\sqrt{3/5}) + \frac{8}{9} \cdot f(0) + \frac{5}{9} \cdot f(\sqrt{3/5})$$

$$\approx 1.71202024520191$$

• The n=4 approximation:

$$\int_{-1}^{1} e^{-\frac{x^2}{2}} dx \approx c_1 f(x_1) + c_2 f(x_2) + c_3 f(x_3) + c_4 f(x_4)$$

$$\approx 1.71122450459949$$

Conclusions

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Gaussian Quadrature (Degree of precision)

Definition (Degree of precision)

The **degree of precision** of a numerical integration method is the greatest integer k for which all degree k or less polynomials are integrated exactly by the method.



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Gaussian Quadrature (Degree of precision)

Example (The degree of precision of the Trapezoid Rule)

$$\int_{a}^{b} f(x) dx = \frac{h}{2} [f(a) + f(b)] - \frac{h^{3}}{12} f''(c)$$

where h = b - a and $a \le c \le b$.

- $f_1(x)$: a polynomial of degree 1 or less;
- $f_2(x) = x^2$.
- $P_i = \frac{h}{2}(f_i(a) + f_i(b))$, $E_i = -\frac{h^3}{12}f_i'(c)$ for i = 1, 2;
- $I_i = \int_a^b f_i(x) dx$ for i = 1, 2.

| \overline{i} | $f_i''(c)$ | E_i | |
|----------------|------------|----------|----------------|
| 1 | 0 | 0 | $P_i = I_i$ |
| 2 | $\neq 0$ | $\neq 0$ | $P_i \neq I_i$ |



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Gaussian Quadrature (Degree of precision)

Newton-Cotes Methods of degree n have degree of precision n (for n odd) and n+1 (for n even).

- The Trapezoid Rule (n = 1) has degree of precision one.
- The Simpson's Rule (n = 2) has degree of precision three.



Conclusions



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Theorem (Degree of precision of Gaussian Quadrature Method)

The Gaussian Quadrature Method, using the degree n Legendre polynomial on [-1,1], has degree of precision 2n-1.

- P(x): the integrand, a polynomial of degree at most 2n-1.
- $p_n(x)$: the *n*th Legendre polynomial.
- x_1, \dots, x_n : n distinct roots in [-1, 1] of $p_n(x)$.





Proof of Theorem

Proof.

We proof that P(x) is integrated exactly by Gaussian Quadrature.

• Using long division of polynomials, we can express:

$$P(x) = S(x)p_n(x) + R(x)$$

, where S(x) and R(x) are polynomials of degree less than n.



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Proof of Theorem

Proof.

We proof that P(x) is integrated exactly by Gaussian Quadrature.

• Using long division of polynomials, we can express:

$$P(x) = S(x)p_n(x) + R(x)$$

, where S(x) and R(x) are polynomials of degree less than n.

- Gaussian Quadrature approximation for P(x) is identical with the integration of R(x) on [-1,1].
 - Gaussian Quadrature is exact on the polynomial R(x):

$$\int_{-1}^1 R(x) dx = \sum_{i=1}^n c_i R(x_i), \text{ where } c_i = \int_{-1}^1 L_i(x) dx, i = 1, \cdots, n$$



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Proof of Theorem

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• Using long division of polynomials, we can express:

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 - Gaussian Quadrature is exact on the polynomial R(x):

$$\int_{-1}^1 R(x) dx = \sum_{i=1}^n c_i R(x_i), \text{ where } c_i = \int_{-1}^1 L_i(x) dx, i = 1, \cdots, n$$

• As $p_n(x_i)=0$ for all i, we get that $P(x_i)=R(x_i)$. So, $\sum_{i=1}^n c_i P(x_i) = \sum_{i=1}^n c_i R(x_i) = \int_{-1}^1 R(x) dx$



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Proof.

Motivation

• The integration for P(x) and R(x) on [-1,1] are identical. By the fact that S(x) is **orthogonal** to $p_n(x)$, we get that:

$$\int_{-1}^{1} P(x) dx = \int_{-1}^{1} S(x) p_n(x) dx + \int_{-1}^{1} R(x) dx = \frac{0}{1} + \int_{-1}^{1} R(x) dx$$

Therefore, it holds that $\int_{-1}^{1} P(x) dx = \sum_{i=1}^{n} c_i P(x_i)$.



Conclusions



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Gaussian Quadrature approximations on a general interval

- $p_n(t)$: the *n*th Legendre polynomial.
- t_1, \dots, t_n : n distinct roots in [-1, 1] of $p_n(t)$.

Using the substitution $t = \frac{(2x-a-b)}{(b-a)}$, we obtain that:

$$\int_{a}^{b} f(x) dx = \int_{-1}^{1} f(\frac{(b-a)t+b+a}{2}) \frac{b-a}{2} dt$$

$$\approx \frac{b-a}{2} \sum_{i=1}^{n} c_{i} f(\frac{(b-a)t_{i}+b+a}{2})$$

, where
$$c_i=\int_{-1}^1 rac{(t-t_1)\cdots\overline{(t-t_i)}(t-t_n)}{(t_i-t_1)\cdots\overline{(t_i-t_i)}(t_i-t_n)}\,dt,\;i=1,\cdots,n.$$



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Gaussian Quadrature approximations on a general interval

Example

Approximate the integral $\int_{1}^{2} \ln x dx$, using Gaussian Quadrature.

- Exact value: $\int_{1}^{2} \ln x dx = 2 \ln 2 1 \ln 1 1 \approx 0.38629436111989$.
- Using the substitution $t = \frac{(2x-1-2)}{(2-1)} = 2x-3$, we obtain that: $\int_1^2 \ln x dx = \int_{-1}^1 \ln(\frac{t+3}{2}) \frac{1}{2} dt$.
- The n=4 approximation:

$$\int_{-1}^{1} \ln(\frac{t+3}{2}) \frac{1}{2} dt \approx c_1 f(t_1) + c_2 f(t_2) + c_3 f(t_3) + c_4 f(t_4)$$

$$\approx 0.38629449693871$$



Conclusions

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Example

Approximate the integral $\int_{1}^{2} \ln x dx$, using Gaussian Quadrature.

- Exact value: $\int_{1}^{2} \ln x dx = 2 \ln 2 1 \ln 1 1 \approx 0.38629436111989$.
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- The n=4 approximation:

$$\int_{-1}^{1} \ln(\frac{t+3}{2}) \frac{1}{2} dt \approx c_1 f(t_1) + c_2 f(t_2) + c_3 f(t_3) + c_4 f(t_4)$$

$$\approx 0.38629449693871$$

- 4-panel composite Trapezoid Rule approximation: ≈ 0.38369950940944 ;
- 4-panel Romberg Integration: $R_{33} \approx 0.38628789352451$.



Conclusions

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Motivation Numerical differentiation Numerical Integration Conclusions

Outline

- Motivation
- 2 Numerical differentiation
 - Finite difference formulas
 - Extrapolation
- 3 Numerical Integration
 - Trapezoid rule
 - Simpson's rule
 - Composite Newton-Cotes formulas
 - Open Newton-Cotes Methods
 - Romberg Integration
 - Adaptive Quadrature
 - Gaussian Quadrature
- 4 Conclusions





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Conclusions

Two methods for Numerical Differentiation

- Finite difference formulas
 - Two-point forward-difference formula
 - Three-point centered-difference formula
 - Three-point centered-difference formula for second derivative
- 2 Extrapolation





Motivation Numerical differentiation Numerical Integration Conclusions

Conclusions

Four methods for Numerical Integration

- Newton-Cotes Formulas for Numerical Integration
 - Closed Newton-Cotes Method: Trapezoid and Simpson's Rules
 - Open Newton-Cotes Method: Midpoint Rule
 - Composite Newton-Cotes Method: Composite Trapezoid (resp. Simpson's / Midpoint) Rule
- 2 Romberg Integration
- Adaptive Quadrature: Adaptive Trapezoid and Simpson's) Rule Quadrature
- Gaussian Quadrature: Gauss-Legendre Quadrature formula



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Thank you!



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