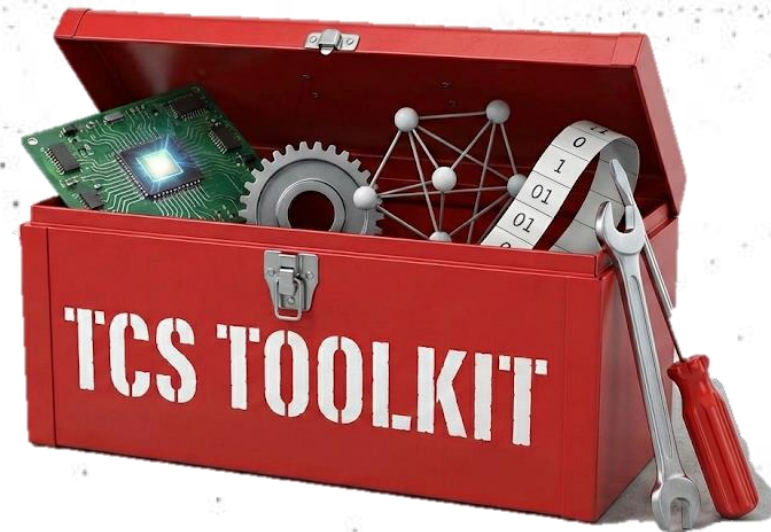


CS 58500 – Theoretical Computer Science Toolkit

Lecture 1 (01/21) Mathematical Basics

https://ruizhezhang.com/course_spring_2026.html



Today's Lecture

- Mathematical Inequalities
 - Taylor approximation
 - Jensen inequality
- Integration
- Stirling Approximation

Mathematical Inequalities: Taylor approximation

Taylor series

For any real or complex-valued function $f(x)$, if it is infinitely differentiable at $x = a$, then it has the following Taylor series expansion:

$$f(x) = f(a) + \frac{f'(a)}{1!}(x-a) + \frac{f''(a)}{2!}(x-a)^2 + \dots = \sum_{n=0}^{\infty} \frac{f^{(n)}(a)}{n!}(x-a)^n$$

Examples:

- $e^x = 1 + x + \frac{x^2}{2} + \frac{x^3}{3!} + \dots$
- $\sin(x) = x - \frac{x^3}{3} + \dots$
- $\ln(1-x) = -x - \frac{x^2}{2} - \frac{x^3}{3} - \dots$

Can we truncate at the k -th term
and control the error?

Mathematical Inequalities: Taylor approximation

Lagrange Form of the Taylor Remainder Theorem

For any real or complex-valued function $f(x)$, under certain continuous condition around a , we have:

$$f(x) = \sum_{n=0}^k \frac{f^{(n)}(a)}{n!} (x-a)^n + \underbrace{\frac{f^{(k+1)}(\xi)}{(k+1)!} (x-a)^{k+1}}_{\text{Reminder } R_k(x)} \quad \xi \in [x, a]$$

Examples:

- $\exp(\pm\epsilon) = 1 \pm \epsilon + \frac{\epsilon^2}{2} + \dots + (-1)^k \frac{\epsilon^k}{k!} + \mathcal{O}(\epsilon^{k+1})$
- $\sin(\epsilon) = \epsilon + \mathcal{O}(\epsilon^3)$
- $\ln(1 - \epsilon) = -\epsilon - \frac{\epsilon^2}{2} + \mathcal{O}(\epsilon^3)$

$$\sin(\epsilon) = 0 + \epsilon + 0 + R_2(\epsilon) = \epsilon - \frac{\cos(\xi)}{3!} \epsilon^3$$

- Since $|\cos(\xi)| \leq 1$, we have
 $\sin(\epsilon) = \epsilon + \mathcal{O}(\epsilon^3)$
- Moreover, for small $\epsilon > 0$, $R_2(\epsilon) < 0$. Thus
 $\sin(\epsilon) < \epsilon$

Mathematical Inequalities: Taylor approximation

Lagrange Form of the Taylor Remainder Theorem

For any real or complex-valued function $f(x)$, under certain continuous condition around a , we have:

$$f(x) = \sum_{n=0}^k \frac{f^{(n)}(a)}{n!} (x-a)^n + \underbrace{\frac{f^{(k+1)}(\xi)}{(k+1)!} (x-a)^{k+1}}_{\text{Reminder } R_k(x)} \quad \xi \in [x, a]$$

Examples:

- $\exp(\pm\epsilon) = 1 \pm \epsilon + \frac{\epsilon^2}{2} + \dots + (-1)^k \frac{\epsilon^k}{k!} + \mathcal{O}(\epsilon^{k+1})$
- $\sin(\epsilon) = \epsilon + \mathcal{O}(\epsilon^3)$
- $\ln(1 - \epsilon) = -\epsilon - \frac{\epsilon^2}{2} + \mathcal{O}(\epsilon^3)$

Trick 1: based on the sign of $R_k(x)$, you can get lower or upper bounds:

- $\exp(-\epsilon) \geq 1 - \epsilon$
- $\ln(1 - \epsilon) \leq -\epsilon$

Trick 2: [Desmos](#)

Mathematical Basics: Convex function and Jensen's inequality

Convex function

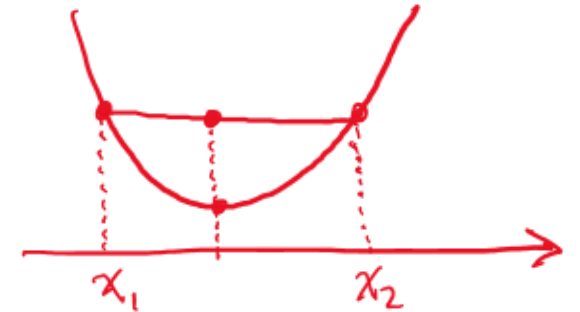
Let f be a real-valued function. Then, f is convex in $[a, b]$ if one of the equivalent conditions holds:

- $\forall x \in [a, b], f^{(2)}(x) \geq 0$
- $\forall \lambda \in [0, 1], x_1, x_2 \in [a, b], f(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda f(x_1) + (1 - \lambda)f(x_2)$

Jensen's inequality

For a convex f , $f\left(\frac{x+y}{2}\right) \leq \frac{f(x)+f(y)}{2}$, “=” if and only if $x = y$

More generally, $f(\mathbb{E}[X]) \leq \mathbb{E}[f(X)]$ for any random variable



Mathematical Basics: Convex function and Jensen's inequality

Jensen's inequality

$$f(\mathbb{E}[X]) \leq \mathbb{E}[f(X)] \quad \forall \text{ convex } f$$

- **AM-GM inequality:**

$$\frac{a+b}{2} \geq \sqrt{ab}, \quad \forall a, b > 0$$

➤ $f(x) = -\ln x$

- **Young's Inequality:**

$$ab \leq \frac{a^p}{p} + \frac{b^q}{q}, \quad \forall a, b > 0, \quad p, q > 1, \quad \frac{1}{p} + \frac{1}{q} = 1$$

➤ $f(x) = -\ln(x)$

Mathematical Basics: Convex function and Jensen's inequality

Jensen's inequality

$$f(\mathbb{E}[X]) \leq \mathbb{E}[f(X)] \quad \forall \text{ convex } f$$

- Cauchy-Schwartz inequality:

$$\begin{aligned} |a_1 b_1 + \dots + a_n b_n| &\leq \left(\sum_i a_i^2 \right)^{\frac{1}{2}} \left(\sum_i b_i^2 \right)^{\frac{1}{2}} \\ \Leftrightarrow \left(\sum_i p_i \frac{a_i}{b_i} \right)^2 &\leq \sum_i p_i \left(\frac{a_i}{b_i} \right)^2, \quad p_i := \frac{b_i^2}{\sum_i b_i^2} \quad \forall i \in [n] \end{aligned}$$

➤ $f(x) = x^2$

- Hölder's inequality:

$$\sum_i |a_i b_i| \leq \left(\sum_i |a_i|^p \right)^{\frac{1}{p}} \left(\sum_i |b_i|^q \right)^{\frac{1}{q}}, \quad \forall p, q > 1, \frac{1}{p} + \frac{1}{q} = 1$$

Mathematical Basics: Convex function and Jensen's inequality

Jensen's inequality

$$f(\mathbb{E}[X]) \leq \mathbb{E}[f(X)] \quad \forall \text{ convex } f$$

- Minkowski's inequality:

$$\left(\sum_i |a_i + b_i|^p \right)^{\frac{1}{p}} \leq \left(\sum_i |a_i|^p \right)^{\frac{1}{p}} + \left(\sum_i |b_i|^p \right)^{\frac{1}{p}} \quad \forall p \geq 1$$

- $f(x) = x^p$

Trick 3: *Inequalities cheat sheet*

https://www.lkozma.net/inequalities_cheat_sheet/ineq.pdf

- We first prove that

$$\sum_i (|a_i| + |b_i|)^p \leq \left(\left(\sum_i |a_i|^p \right)^{\frac{1}{p}} + \left(\sum_i |b_i|^p \right)^{\frac{1}{p}} \right)^p = (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p)^p$$

- Observe that

$$\begin{aligned} |a_i| + |b_i| &= (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p) \left(\frac{\|\mathbf{a}\|_p}{\|\mathbf{a}\|_p + \|\mathbf{b}\|_p} \frac{|a_i|}{\|\mathbf{a}\|_p} + \frac{\|\mathbf{b}\|_p}{\|\mathbf{a}\|_p + \|\mathbf{b}\|_p} \frac{|b_i|}{\|\mathbf{b}\|_p} \right) \\ &=: (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p) \left(\lambda \frac{|a_i|}{\|\mathbf{a}\|_p} + (1 - \lambda) \frac{|b_i|}{\|\mathbf{b}\|_p} \right) \end{aligned}$$

- Since x^p is convex for $p \geq 1$, by Jensen's inequality,

$$(|a_i| + |b_i|)^p \leq (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p)^p \left(\lambda \frac{|a_i|^p}{\|\mathbf{a}\|_p^p} + (1 - \lambda) \frac{|b_i|^p}{\|\mathbf{b}\|_p^p} \right)$$

- Summing over i :

$$\sum_i (|a_i| + |b_i|)^p \leq (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p)^p \left(\lambda \sum_i \frac{|a_i|^p}{\|\mathbf{a}\|_p^p} + (1 - \lambda) \sum_i \frac{|b_i|^p}{\|\mathbf{b}\|_p^p} \right)$$

- Summing over i :

$$\begin{aligned}
 \sum_i (|a_i| + |b_i|)^p &\leq (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p)^p \left(\lambda \sum_i \frac{|a_i|^{\color{red}p}}{\|\mathbf{a}\|_p^{\color{red}p}} + (1 - \lambda) \sum_i \frac{|b_i|^{\color{red}p}}{\|\mathbf{b}\|_p^{\color{red}p}} \right) \\
 &= (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p)^p \left(\lambda \frac{\|\mathbf{a}\|_p^{\color{red}p}}{\|\mathbf{a}\|_p^{\color{red}p}} + (1 - \lambda) \frac{\|\mathbf{b}\|_p^{\color{red}p}}{\|\mathbf{b}\|_p^{\color{red}p}} \right) \\
 &= (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p)^p
 \end{aligned}$$

- By triangle inequality,

$$|a_i + b_i| \leq |a_i| + |b_i|$$

- Thus, we have

$$\sum_i |a_i + b_i|^p \leq \sum_i (|a_i| + |b_i|)^p \leq (\|\mathbf{a}\|_p + \|\mathbf{b}\|_p)^p$$



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- Mathematical Inequalities
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- Integration
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Integration

Numerical integration

Given **query access** to $f: [a, b] \rightarrow \mathbb{R}$, how to estimate

$$\int_a^b f(x) dx$$



Meta algorithm: Quadrature

1. Choose a set of points and coefficients $\{(x_i, w_i)\}_{i \in [n]}$
2. Output

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

Integration

Midpoint rule / Riemann sum

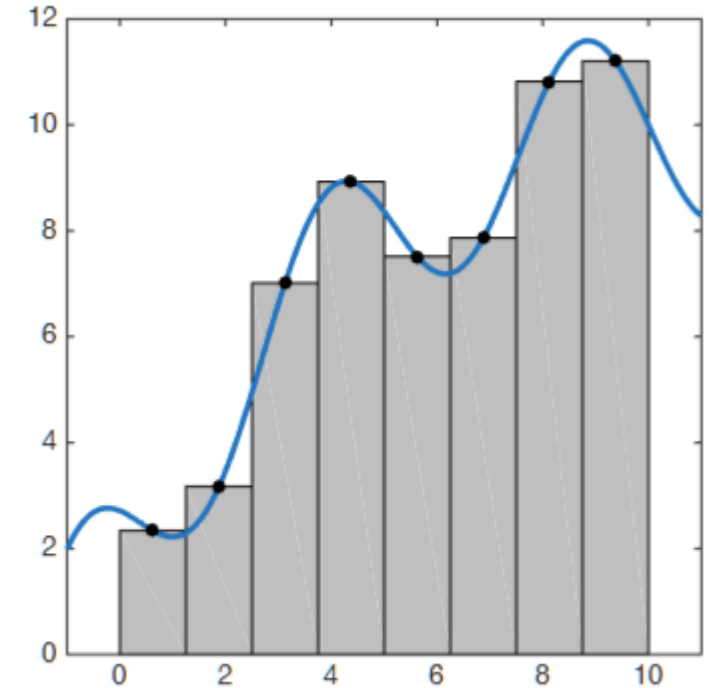
$$\int_0^1 f(x) dx \approx \frac{1}{n} \sum_{i=0}^{n-1} f\left(\frac{i+0.5}{n}\right)$$

What is the approximation error?

Theorem (Midpoint rule error bound).

Suppose that $|f''(x)| \leq M_2$ for $x \in [0,1]$. Then, we have

$$\left| \int_0^1 f(x) dx - \frac{1}{n} \sum_{i=0}^{n-1} f\left(\frac{i+0.5}{n}\right) \right| \leq \frac{M_2}{24n^2}$$



Midpoint rule error bound proof

- Consider the first interval $\left[0, \frac{1}{n}\right]$. By Taylor's theorem,

$$f(x) = f\left(\frac{1}{2n}\right) + f'\left(\frac{1}{2n}\right)\left(x - \frac{1}{2n}\right) + f''(\xi_x)\left(x - \frac{1}{2n}\right)^2$$

- Integration:

$$\int_0^{\frac{1}{n}} f(x) dx = \frac{1}{n} f\left(\frac{1}{2n}\right) + 0 + \int_0^{\frac{1}{n}} f''(\xi_x) \left(x - \frac{1}{2n}\right)^2 dx$$

$$\|f - M_2\| \leq M_2 \text{ for } x \in [0, 1]$$

Midpoint rule error bound proof

- Consider the first interval $\left[0, \frac{1}{n}\right]$. By Taylor's theorem,

$$f(x) = f\left(\frac{1}{2n}\right) + f'\left(\frac{1}{2n}\right)\left(x - \frac{1}{2n}\right) + f''(\xi_x)\left(x - \frac{1}{2n}\right)^2$$

- Integration:

$$\begin{aligned}\int_0^{\frac{1}{n}} f(x) dx &= \frac{1}{n} f\left(\frac{1}{2n}\right) + 0 + \int_0^{\frac{1}{n}} f''(\xi_x) \left(x - \frac{1}{2n}\right)^2 dx \\ &= \frac{1}{n} f\left(\frac{1}{2n}\right) \pm M_2 \int_0^{\frac{1}{n}} \left(x - \frac{1}{2n}\right)^2 dx = \frac{1}{n} f\left(\frac{1}{2n}\right) \pm \frac{M_2}{24} n^{-3}\end{aligned}$$

if $|f''(x)| \leq M_2$ for $x \in [0, 1]$

- Summing over n intervals together proves the theorem.

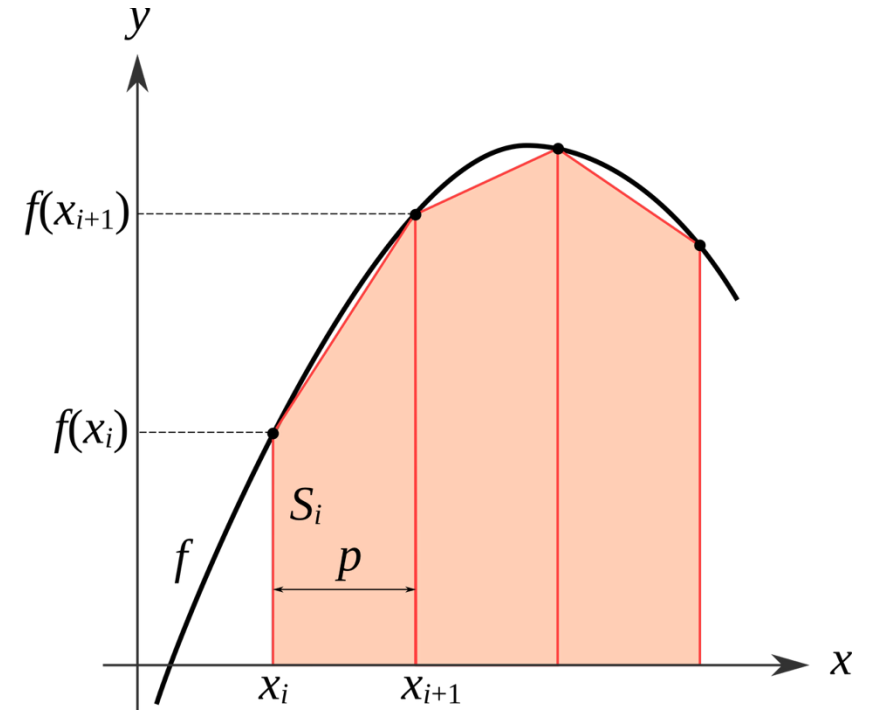


Integration

Trapezoidal rule

$$\begin{aligned}\int_0^1 f(x) dx &\approx \sum_{i=0}^{n-1} \frac{1}{2} \left(f\left(\frac{i}{n}\right) + f\left(\frac{i+1}{n}\right) \right) \frac{1}{n} \\ &= \frac{1}{2n} \left(f(0) + 2 \sum_{i=1}^{n-1} f\left(\frac{i}{n}\right) + f(1) \right)\end{aligned}$$

- Error bound: $\mathcal{O}(M_2 n^{-2})$



Integration

Interpolatory quadrature / Gaussian quadrature

$$\int_0^1 f(x) dx \approx \int_0^1 p(x) dx = \int_0^1 \sum_{i=0}^n f(x_i) \ell_i(x) dx = \sum_{i=0}^n f(x_i) \underbrace{\int_0^1 \ell_i(x) dx}_{w_i}$$

Lagrange interpolation

For any f , there is a unique degree- n polynomial p such that $f(x_i) = p(x_i)$ for any given $n + 1$ points x_0, \dots, x_n . More specifically,

$$p(x) = \sum_{i=0}^n f(x_i) \ell_i(x), \quad \ell_i(x) := \prod_{0 \leq j \leq n, j \neq i} \frac{x - x_j}{x_i - x_j}$$

Integration

Interpolatory quadrature / Gaussian quadrature

$$\int_0^1 f(x) dx \approx \int_0^1 p(x) dx = \int_0^1 \sum_{i=0}^n f(x_i) \ell_i(x) dx = \sum_{i=0}^n f(x_i) \underbrace{\int_0^1 \ell_i(x) dx}_{w_i}$$

- If f is a polynomial of degree $\leq n$, then there is no error!
- Suppose we have an interpolation scheme such that

$$\left| \int_0^1 f(x) dx - \sum_{i=0}^n w_i f(x_i) \right| \leq C_n M_{n+1}, \quad \text{where } M_{n+1} := \max_{x \in [0,1]} |f^{(n+1)}(x)|$$

Then, we can cut the integral into k pieces and get the following

$$\left| \int_0^1 f(x) dx - \frac{1}{k} \sum_{j=0}^{k-1} \sum_{i=0}^n w_i f\left(\frac{x_i + j}{k}\right) \right| \leq k \cdot C_n \frac{M_{n+1}}{k^{n+2}} = C_n \frac{M_{n+1}}{k^{n+1}} \quad (\text{exp convergence!})$$

Apply chain rule $n + 1$ times

Integration

Interpolatory quadrature / Gaussian quadrature

$$\left| \int_0^1 f(x) dx - \sum_{i=0}^n w_i f(x_i) \right| \leq C_n M_{n+1}, \quad \text{where } M_{n+1} := \max_{x \in [0,1]} |f^{(n+1)}(x)|$$

Lemma. Given any $n + 1$ times differentiable function f . Let $p_n(x)$ be an n degree polynomial such that $p_n(x_i) = f(x_i)$ for $i = 0, 1, \dots, n$. Then, for any $x \in [0, 1]$, we have

$$f(x) - p_n(x) = \frac{f^{(n+1)}(\xi_x)}{(n+1)!} \prod_{i=0}^n (x - x_i) \quad \text{for some } \xi_x \in [0, 1]$$

- This lemma implies that

$$\left| \int_0^1 f(x) dx - \sum_{i=0}^n w_i f(x_i) \right| \leq \left| \int_0^1 \frac{f^{(n+1)}(\xi_x)}{(n+1)!} \prod_{i=0}^n (x - x_i) dx \right| \leq \frac{M_{n+1}}{(n+1)!} \cdot \max_{x \in [0,1]} \left| \prod_{i=0}^n (x - x_i) \right|$$

Integration

Lemma. Given any $n + 1$ times differentiable function f . Let $p_n(x)$ be an n degree polynomial such that $p_n(x_i) = f(x_i)$ for $i = 0, 1, \dots, n$. Then, for any $x \in [0, 1]$, we have

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Proof.

- Let $\omega(t) := \prod_{i=0}^n (t - x_i)$ and consider the function

$$F(t) := f(t) - p_n(t) - \frac{f(x) - p_n(x)}{\omega(x)} \omega(t)$$

- $F(x_i) = 0$ for $i = 0, 1, \dots, n$ and $F(x) = 0$ (check by yourself)
- $F(t)$ has $n + 2$ roots in $[0, 1]$, by applying [Rolle's Theorem](#) repeatedly, we get that there exists $\xi \in [0, 1]$ such that $F^{(n+1)}(\xi) = 0$

$$F^{(n+1)}(\xi) = f^{(n+1)}(\xi) - 0 - \frac{f(x) - p_n(x)}{\omega(x)} = 0$$



Integration

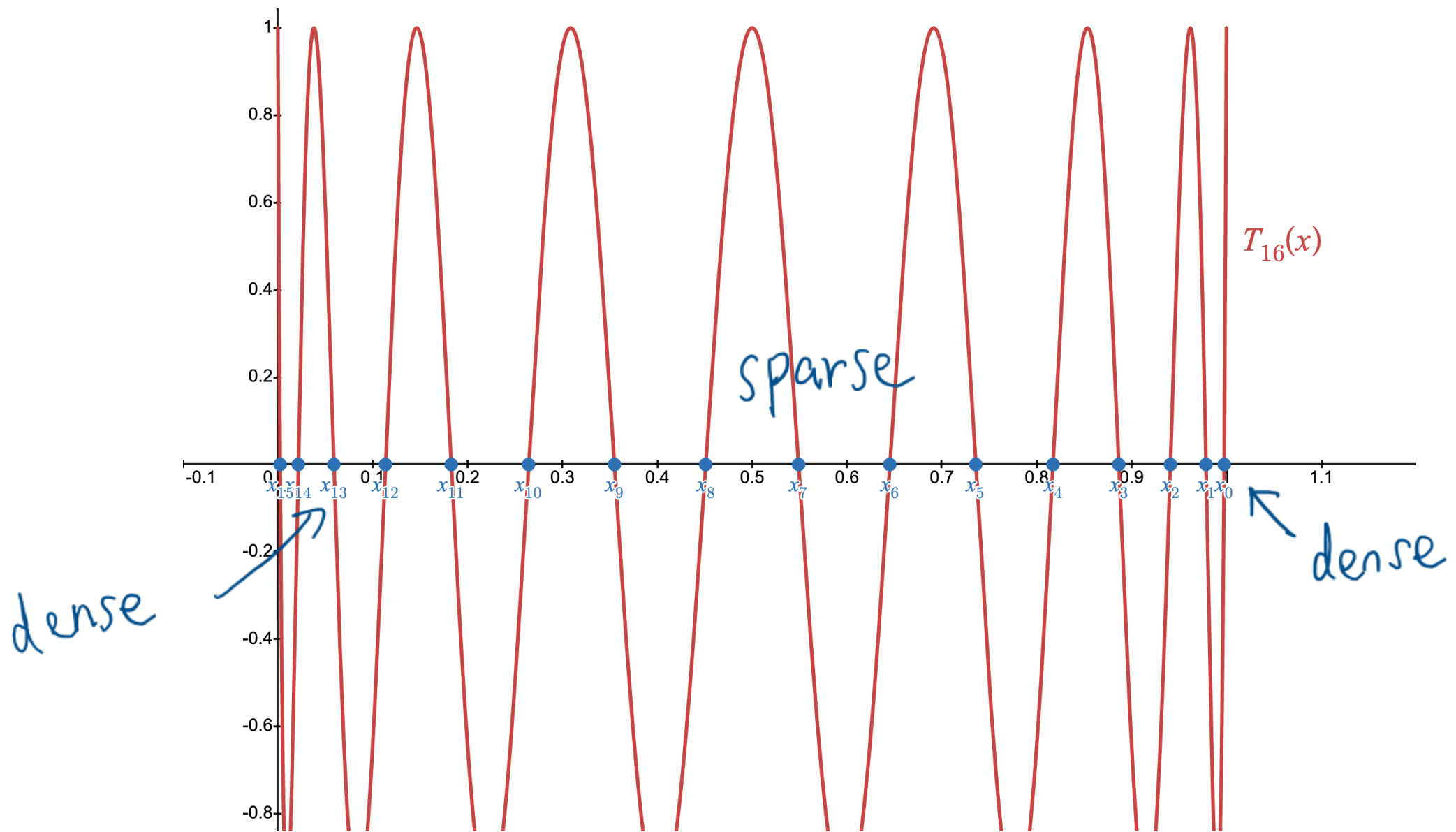
Interpolatory quadrature / Gaussian quadrature

$$\left| \int_0^1 f(x) dx - \sum_{i=0}^n w_i f(x_i) \right| \leq C_n M_{n+1}, \quad \text{where } M_{n+1} := \max_{x \in [0,1]} |f^{(n+1)}(x)|$$

- How to choose $x_0, x_1, \dots, x_n \in [0,1]$ such that $\max_{x \in [0,1]} |\prod_{i=0}^n (x - x_i)|$ is small?
- This quantity is minimized by Chebyshev nodes:

$$x_i := \frac{1}{2} + \frac{1}{2} \cos \left(\frac{2i+1}{2n+2} \pi \right) \quad \forall i = 0, 1, \dots, n$$

- $\{x_i\}$ are the roots of the Chebyshev polynomial $T_{n+1}(2x - 1)$



<https://www.desmos.com/calculator/dnyqzmwkei>

Integration

Interpolatory quadrature / Gaussian quadrature

$$\left| \int_0^1 f(x) dx - \sum_{i=0}^n w_i f(x_i) \right| \leq C_n M_{n+1}, \quad \text{where } M_{n+1} := \max_{x \in [0,1]} |f^{(n+1)}(x)|$$

- How to choose $x_0, x_1, \dots, x_n \in [0,1]$ such that $\max_{x \in [0,1]} |\prod_{i=0}^n (x - x_i)|$ is small?
- This quantity is minimized by Chebyshev nodes:

$$x_i := \frac{1}{2} + \frac{1}{2} \cos \left(\frac{2i+1}{2n+2} \pi \right) \quad \forall i = 0, 1, \dots, n$$

- We have that

$$\left| \int_0^1 f(x) dx - \sum_{i=0}^n w_i f(x_i) \right| \leq \frac{M_{n+1}}{(n+1)! 2^{2n+1}}$$

Integration

Exponentially Convergent Trapezoidal Rule

- If f is sufficiently smooth and periodic, then the trapezoidal rule / midpoint rule has an exponentially convergent error bound.
- Exponential convergence also holds for some “peak-like” functions integrated over the real line, e.g.,

$$\underbrace{\left| \int_{-\infty}^{\infty} e^{-x^2} dx - h \sum_{j=-\infty}^{\infty} e^{-(jh)^2} \right|}_{\sqrt{\pi}} = \mathcal{O}(e^{-\pi^2/h^2})$$

Exponentially Convergent Trapezoidal Rule

- Let $I_h := h \sum_{j=-\infty}^{\infty} e^{-(jh)^2}$ and $I = \int_{-\infty}^{\infty} e^{-x^2} dx$

- Consider the **Fourier transform** of $f(x) = e^{-x^2}$:

$$\hat{f}(\xi) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-i\xi x} e^{-x^2} dx = \frac{1}{2\sqrt{\pi}} e^{-\xi^2/4}$$

- Observe that $I = 2\pi \hat{f}(0) = \sqrt{\pi}$
- Another tool we use is the **Poisson Summation Formula**, which connects discrete sum I_h to the continuous Fourier transform \hat{f} :

$$I_h = 2\pi \sum_{j=-\infty}^{\infty} \hat{f}\left(\frac{2\pi j}{h}\right) = \sqrt{\pi} \sum_{j=-\infty}^{\infty} e^{-\pi^2 j^2 / h^2}$$

$$|I - I_h| = 2\sqrt{\pi} \sum_{j=1}^{\infty} e^{-\pi^2 j^2 / h^2}$$

Exponentially Convergent Trapezoidal Rule

$$I = 2\pi\hat{f}(0)$$

$$I_h = 2\pi \sum_{j=-\infty}^{\infty} \hat{f}\left(\frac{2\pi j}{h}\right) = \sqrt{\pi} \sum_{j=-\infty}^{\infty} e^{-\pi^2 j^2 / h^2}$$

- Thus, we have

$$\begin{aligned} |I - I_h| &= 2\pi \sum_{j \neq 0} \hat{f}\left(\frac{2\pi j}{h}\right) = 2\sqrt{\pi} \sum_{j=1}^{\infty} e^{-\pi^2 j^2 / h^2} \\ &\leq 2\sqrt{\pi} \sum_{j=1}^{\infty} (e^{-\pi^2 / h^2})^j = 2\sqrt{\pi} e^{-\pi^2 / h^2} \frac{1}{1 - e^{-\pi^2 / h^2}} = \mathcal{O}(e^{-\pi^2 / h^2}) \end{aligned}$$

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Stirling Approximation

- $\ln(n!) = \sum_{i=1}^n \ln(i)$
- Trapezoidal rule:

$$\int_1^n \ln(x) dx \stackrel{\geq}{\approx} \frac{1}{2} (\ln(1) + \ln(n)) + \sum_{i=2}^{n-1} \ln(i) = \ln(n!) - \frac{1}{2} \ln(n)$$

- Evaluate the integral and simplify the terms:

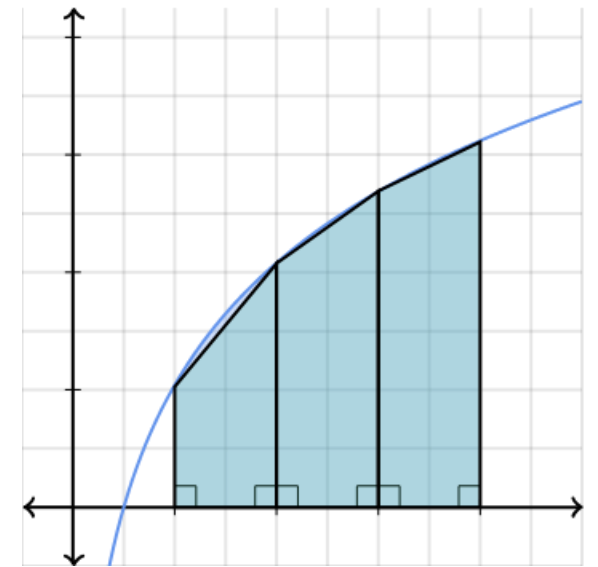
$$\int_1^n \ln(x) dx = x \ln(x) - x \Big|_1^n = n \ln(n) - n + 1$$

$$\ln(n!) \approx \left(n + \frac{1}{2}\right) \ln(n) - n + 1 \quad \Rightarrow \quad n! \approx e^{\sqrt{n}} \left(\frac{n}{e}\right)^n$$

\leq

$$n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$

concave
increasing



Stirling Approximation

- Trapezoidal rule:

$$\int_1^n \ln(x) dx = \ln(n!) - \frac{1}{2} \ln(n) + E_n, \quad (E_n > 0)$$

- Evaluate the integral and simplify the terms:

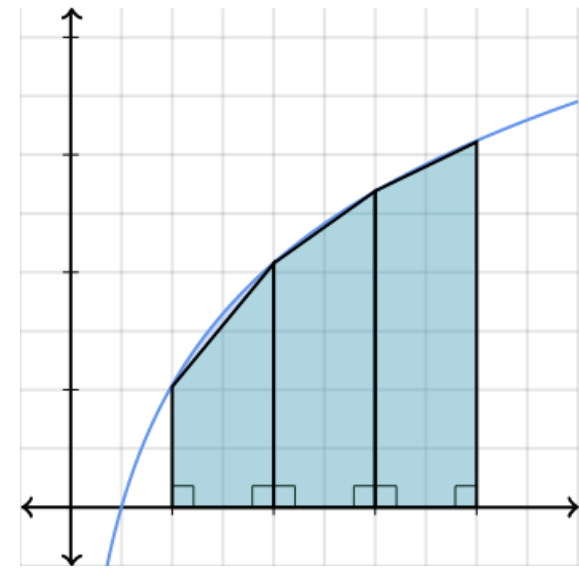
$$\ln(n!) = \left(n + \frac{1}{2}\right) \ln(n) - n + 1 - E_n \quad \Rightarrow \quad n! = e^{1-E_n} \sqrt{n} \left(\frac{n}{e}\right)^n$$

- For each segment $[i, i + 1]$,

$$\int_i^{i+1} \ln(x) dx - \frac{1}{2} (\ln(i) + \ln(i + 1)) = -\frac{(\ln(x))''|_{x=\xi_i}}{12} = \frac{1}{12\xi_i^2}, \quad \xi_k \in (i, i + 1)$$

- Summing together,

$$E_n = \sum_{i=1}^{n-1} \frac{1}{12\xi_i^2} < \sum_{i=1}^{n-1} \frac{1}{12i^2} = \mathcal{O}(1), \quad E_n > \sum_{i=1}^{n-1} \frac{1}{12(i+1)^2} = \Omega(1) \quad \Rightarrow \quad E_n = C + o(1)$$



Stirling Approximation

$$n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \mathcal{O}\left(\frac{1}{n}\right)\right)$$

Asymptotic estimation of binomial coefficient

- Basic version:

$$\frac{n^k}{k^k} \leq \binom{n}{k} < \left(\frac{en}{k}\right)^k$$

- Entropy version:

$$\frac{2^{nH(p)}}{\sqrt{8p(1-p)n}} \leq \binom{n}{pn} \leq \frac{2^{nH(p)}}{\sqrt{2\pi p(1-p)n}}$$

where $H(p) := -p \log_2(p) - (1-p) \log_2(1-p)$ for any $p \in (0,1)$

Binormal Coefficient Estimation

$$\frac{2^{nH(p)}}{\sqrt{8p(1-p)n}} \leq \binom{n}{pn} \leq \frac{2^{nH(p)}}{\sqrt{2\pi p(1-p)n}}$$

- By Stirling approximation,

$$\begin{aligned} \binom{n}{pn} &\approx \frac{\sqrt{2\pi n} \left(\frac{n}{e}\right)^n}{\sqrt{2\pi pn} \left(\frac{pn}{e}\right)^{pn} \cdot \sqrt{2\pi(1-p)n} \left(\frac{(1-p)n}{e}\right)^{(1-p)n}} \\ &= \frac{1}{\sqrt{2\pi p(1-p)n}} \cdot e^{pn+(1-p)n-n} \cdot 2^{n(\log n - p \log(pn) - (1-p) \log((1-p)n))} \\ &= \frac{1}{\sqrt{2\pi p(1-p)n}} \cdot 2^{n(-p \log p - (1-p) \log(1-p))} = \frac{2^{nH(p)}}{\sqrt{2\pi p(1-p)n}} \end{aligned}$$