



SUPPLEMENTAL MATHEMATICAL FACTS

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A.1 Basic logic for proof

[1, p. 60]The negation of

for any $\epsilon > 0$, there exist $N > 0$, such that for all $n > N$, we have $|a_n - a| < \epsilon$

is

there exist an $\epsilon > 0$, such that for every $N > 0$, such that for all $n > N$, we have $|a_n - a| > \epsilon$.

[1, p. 60]The negation of

for any $\epsilon > 0$, there exist $\delta > 0$, such that for all $|x - x_0| < \delta$, we have $|f(x) - f(x_0)| < \epsilon$

is

there exist an $\epsilon > 0$, such that for every $\delta > 0$, such that for all $|x - x_0| < \delta$, we have $|f(x) - f(x_0)| > \epsilon$.

A.2 Some common limits

Lemma A.2.1 (Stirling approximation). • For positive integer n ,

$$\ln n! = n \ln n - n + O(\ln n).$$

•

$$\sqrt{2\pi n} n^{n+1/2} e^{-n} \leq n! \leq e n^{n+1/2} e^{-n}, \forall n > 0.$$

Lemma A.2.2 (common limits summary).

•

$$\lim_{n \rightarrow \infty} \frac{\ln n}{n} = 0.$$

•

$$\lim_{n \rightarrow \infty} \frac{x^n}{n!} = 0, \forall x \in \mathbb{R}.$$

•

$$\lim_{n \rightarrow \infty} \frac{n!}{n^n} = 0.$$

•

$$\lim_{n \rightarrow \infty} M^{1/n} = 1$$

for any $M > 0$.

•

$$\lim_{n \rightarrow \infty} \frac{\ln n!}{n} = \infty, \lim_{n \rightarrow \infty} (n!)^{1/n} = \infty.$$

Proof. (2) see [Lemma 3.6.4](#) and [Lemma 1.4.3](#). (3) [Lemma 1.4.3](#). (4) [Lemma 1.5.1](#). (5) (a) Use Stirling approximation $\ln n! = n \ln n - n + O(\ln n)$ and $\ln n!/n = n - 1 + O(\ln n/n) \rightarrow \infty$. (b) Note that $(n!)^{1/n} = \exp(\ln(n!)^{1/n}) = \exp(\frac{\ln n!}{n})$. \square

Note A.2.1. A helpful and general summary, as $n \rightarrow \infty$

$$\ln n \ll n^r (r > 0) \ll a^n (a > 1) \ll n! \ll n^n.$$

Lemma A.2.3 (property of e). Define

$$\lim_{n \rightarrow \infty} (1 + 1/n)^n = e$$

and then

$$\lim_{n \rightarrow \infty} (1 + x/n)^n = e^x$$

for any real x .

Proof.

$$\lim_{n \rightarrow \infty} ((1 + x/n)^{n/x})^x = e^x$$

use the fact the $f(y) = y^x$ is continuous, such that function evaluation and limit can be exchanged. □

A.3 Common series summation

Lemma A.3.1. [2, p. 1]

•

$$\sum_{k=1}^n k = \frac{n(n+1)}{2}$$

•

$$\sum_{k=1}^n k^2 = \frac{n(n+1)(2n+1)}{6}$$

•

$$\sum_{k=1}^n k^3 = \left[\frac{n(n+1)}{2} \right]^2$$

Lemma A.3.2. [2, p. 1]

Assume $q \neq 1$.

•

$$\sum_{k=1}^n aq^{k-1} = a \frac{q^n - 1}{q - 1}$$

•

$$\sum_{k=0}^{n-1} kq^k = \frac{(n-1)q^n}{q-1} + \frac{(q-q^n)}{(q-1)^2}$$

•

$$\sum_{k=0}^{n-1} (n-1-k)q^k = -(n-1) \frac{1}{q-1} - \frac{(q-q^n)}{(q-1)^2}$$

Proof. (3) use (1)(2), we have

$$\sum_{k=0}^{n-1} (n-1-k)q^k = (n-1) \frac{q^n - 1}{q-1} - \frac{(n-1)q^n}{q-1} - \frac{(q-q^n)}{(q-1)^2}$$

□

A.4 Some common spaces

The metric space (\mathbb{R}^n, d_2) is the set \mathbb{R}^n with metric $d_2(x, y) = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$.

[3, p. 122] The metric space l^2 is the set of all infinite sequence of real or complex numbers $\{x_1, x_2, \dots\}$ such that $\sum_{i=1}^{\infty} x_i^2 < \infty$, i.e., $\sum_{i=1}^{\infty} x_i^2$ converges. The metric is usually defined as

$$d_2(\{x_n\}, \{y_n\}) = \sqrt{\sum_{k=1}^{\infty} (x_k - y_k)^2}$$

The metric space $l^p, 1 \leq p < \infty$, is the set of all infinite sequence of real or complex numbers $\{x_1, x_2, \dots\}$ such that $\sum_{i=1}^{\infty} |x_i|^p < \infty$, i.e., $\sum_{i=1}^{\infty} |x_i|^p$ converges. The metric is usually defined as

$$d_p(\{x_n\}, \{y_n\}) = \sqrt[p]{\sum_{k=1}^{\infty} (x_k - y_k)^p}$$

The metric space l^{∞} , is the set all infinite sequence of real or complex numbers $\{x_1, x_2, \dots\}$ such that every x_i is bounded. The metric is defined as

$$d_{\infty}(\{x_n\}, \{y_n\}) = \sup_n |x_n - y_n|$$

[4, p. 75]. The metric space $C[a, b] = (C[a, b], d_{\infty})$ denote the set of real-valued (or complex valued) functions defined on the interval $[a, b]$. The metric d_{∞} is given as

$$d_{\infty}(x, y) = \sup_t |x(t) - y(t)|$$

Remark A.4.1. Caution! Sometimes $C[a, b]$ refers to only continuous functions. [5, p. 23]

The metric space $(C[a, b], d_p)$ denote the set of real-valued (or complex valued) functions defined on the interval $[a, b]$. The metric d_p is given as

$$d_p(x, y) = \left[\int_a^b |x(t) - y(t)|^p dt \right]^{1/p}$$

where $1 \leq p < \infty$. [4, p. 75].

The vector space $\mathcal{L}(V, W)$ usually denotes the set of all linear operators from V into W .

A.4.1 Notations on continuously differentiable functions

- C^0 refers to continuous function

- C^1 refers to functions having continuous first derivatives, also called continuously differentiable functions.
- C^2 refers to functions having continuous second derivatives
- C^∞ refers to smooth functions

A.5 Different modes of continuity

Chain of inclusions for functions over a closed and bounded subset of the real line

$$\text{continuouslyDifferentiable} \subseteq \text{LipschitzContinuous} \subseteq \text{UniformlyContinuous}$$

Remark A.5.1.

- Continuously differentiable on a closed interval indicates the derivative is bounded $f' \leq M$, then we have

$$|f(x) - f(y)| = f'(s)|x - y| \leq M|x - y|$$

hence Lipschitz continuous.

- $f(x) = |x|$ is Lipschitz continuous but is not differentiable everywhere except at $x = 0$, therefore it is not continuously differentiable.
- Lipschitz continuous \rightarrow continuous:

$$|f(x) - f(y)| \leq L|x - y| \rightarrow 0$$

as $|x - y| \rightarrow 0$

Lemma A.5.1 (differentiable implies continuous). *If f is differentiable on $[a, b]$, then it is continuous on $[a, b]$.*

Proof:

$$\begin{aligned} \lim_{y \rightarrow x} f(y) - f(x) &= \lim_{y \rightarrow x} (y - x)(f(y) - f(x)) / (y - x) = \\ &= \lim_{y \rightarrow x} (y - x) \lim_{y \rightarrow x} (f(y) - f(x)) / (y - x) = 0 \end{aligned}$$

where we have use the property that if two limits exist then they can multiply.[3, p. 42].

Remark A.5.2. This lemma indicates that a function differentiable everywhere will be continuous everywhere.

Lemma A.5.2 (differentiable everywhere NOT implies continuously differentiable). *A function is differentiable everywhere NOT implies it is continuously differentiable function.*

The standard example is

$$f(x) = \begin{cases} x^2 \sin(\frac{1}{x}), & x \neq 0 \\ 0, & x = 0 \end{cases}$$

This function can be differentiated every where and $f'(0) = 0$, but $\lim_{x \rightarrow 0} f'(x)$ does not exist. See [link](#).

A.5.1 continuity vs. uniform continuity

Definition A.5.1. A function $f : X \rightarrow Y$ is uniformly continuous if for every $\epsilon > 0$ there exist a $\delta > 0$ such that for every $x, x_0 \in X$,

$$\rho(x, x_0) \leq \delta \Rightarrow \rho(f(x), f(x_0)) < \epsilon$$

Theorem A.5.1. [3, p. 154] If f is a continuous function from a compact metric space M_1 into a metric space M_2 , then f is uniformly continuous on M_1 .

Corollary A.5.1.1. [3, p. 154] If f is a continuous real-value function on a closed and bounded subset X of \mathbb{R}^n , then f is uniformly continuous on X .

Example A.5.1. The function $f(x) = x^2$ is continuous but not uniformly continuous on the interval $(0, \infty)$.

Lemma A.5.3 (sufficient condition). Let $S = \mathbb{R}$. if f is global Lipschitz continuous, i.e.

$$|f(x_1) - f(x_2)| < M|x_1 - x_2|$$

$\forall x_1, x_2 \in S$, then f is uniformly continuous.

Proof: $|f(x_1) - f(x_2)| < M|x_1 - x_2| \rightarrow 0$

A.6 Exchanges of limits

A.6.1 Overall remark

Remark A.6.1.

- Usually, the necessary conditions for exchanging limits is difficult to find, therefore only sufficient conditions are given.
- Many operations are in nature taking limits, for example, summing infinite terms is taking limits on partial sums; integrals is taking limits on both summation and partitions; derivative is taking limits on quotient expressions.

A.6.2 exchange limits with infinite summations

Let $\lim_{m \rightarrow \infty} \sum_{n=1}^{\infty} f(m, n) = \sum_{n=1}^{\infty} \lim_{m \rightarrow \infty} f(m, n)$ Based on dominated convergence, if there is a $g(n)$ such that $f(m, n) < g(n), \forall m$ and $\sum_{n=1}^{\infty} g(n)$ exists, then we can exchange.

To use the dominated convergence theorem in Lebesgue integral, we can define a simple function s_n on $[0, \infty]$ take $f(m, n)$ on the interval $[m-1, m)$. Then the integral of s_n with respect to Lebesgue measure on real line will give the $\int_{[0, \infty)} s_n d\mu = \sum_{m=1}^{\infty} f(m, n)$

Theorem A.6.1. [3, pp. 94, 373] Let $a_{m,n}$ be non-negative and $\sum_m \sum_n a_{m,n}$ exists, then

$$\sum_m \sum_n a_{m,n} = \sum_n \sum_m a_{m,n}$$

Corollary A.6.1.1. Let $a_{m,n}$ be increasing on both m, n and $\lim_{m \rightarrow \infty} \lim_{n \rightarrow \infty} a_{m,n}$ exists, then

$$\lim_{m \rightarrow \infty} \lim_{n \rightarrow \infty} a_{m,n} = \lim_{n \rightarrow \infty} \lim_{m \rightarrow \infty} a_{m,n}$$

Proof: by constructing partial sums.

A.6.3 Exchange limits with integration and differentiation

Theorem A.6.2. [3, p. 249] Let α be a function of bounded variation on $[a, b]$ and let f_n be a sequence of functions in $\mathcal{R}_\alpha[a, b]$ which converges uniformly to a function f . Then $f \in \mathcal{R}_\alpha[a, b]$ and

$$\lim_{n \rightarrow \infty} \int_a^b f_n d\alpha = \int_a^b \lim_{n \rightarrow \infty} f d\alpha$$

Theorem A.6.3. [3, p. 249] Let $\{f_n\}$ be a sequence of differentiable functions on (a, b) . Suppose that

- f'_n is continuous on (a, b)
- $\{f_n\}$ converges pointwise to f
- $\{f'_n\}$ converges uniformly

then f is differentiable on (a, b) and f'_n converges uniformly to f' .

A.6.4 Exchange differentiation with integration

Theorem A.6.4. Let $f(x, y)$ be continuous on $[a, b] \times [c, d]$. Then

$$\phi(y) = \int_a^b f(x, y) dx$$

defined above is continuous function on $[c, d]$

Proof: for any $\epsilon > 0$, there exist δ , such that

$$|\phi(y) - \phi(y')| \leq \int_a^b |f(x, y) - f(x, y')| dx \leq \epsilon(b-a) \forall |y - y'| < \delta$$

where we have the fact of $f(x, y) - f(x, y')$ is bounded (since continuous function on a compact set is uniformly continuous and will have maximum and minimum) which shows $\phi(y)$ is uniformly continuous.

Theorem A.6.5. Let f and f_y be continuous on $[a, b] \times [c, d]$. Then ϕ is differentiable and

$$\phi_y = \int_a^b f_y(x, y) dx$$

Proof:

$$\frac{\phi(y+h) - \phi(y)}{h} = \frac{1}{h} \int_a^b f(x, y+h) - f(x, y) dx = \int_a^b f_y(x, z) dz$$

due to Taylor theorem, where $z \in [y, y + h]$. Then

$$\left| \frac{\phi(y + h) - \phi(y)}{h} - \int_a^b f_y(x, y) dx \right| \leq \int_a^b |f_y(x, z) - f_y(x, y)| dx$$

Because f_y is continuous on compact set, then it is uniformly continuous. Therefore given $\epsilon > 0$, there exists δ such that

$$|f_y(x, y') - f_y(x, y)| < \epsilon / (b - a), \forall |y - y'| < \delta$$

Taking $h < \delta$, we have

$$\left| \frac{\phi(y + h) - \phi(y)}{h} - \int_a^b f_y(x, y) dx \right| < \epsilon.$$

Take the limit on h and we get the result.

A.6.5 Exchange limit and function evaluations

Lemma A.6.1. *Let $\{x_n\}$ be a sequence with limit x , let f be a continuous function*

$$\lim_{n \rightarrow \infty} f(x_n) = f(\lim_{n \rightarrow \infty} x_n) = f(x)$$

Proof: from the definition of continuous function.

A.7 Useful inequalities

Lemma A.7.1 (arithmetic-geometric mean inequality). For $x_1, \dots, x_n \geq 0$, we have

$$(x_1 x_2 \dots x_n)^{1/n} \leq \sum_{i=1}^n x_i / n.$$

Specifically,

$$\frac{x_1 + x_2}{2} \geq \sqrt{x_1 x_2}$$

Proof. use $y = \ln(x)$ and concavity of $\ln(x)$ □

A.7.1 Gronwall's inequality

see [\[6\]](#)

A.7.2 Inequality for norms

Lemma A.7.2. [\[7\]](#) For L^p normed space, we have

$$\|x\|_2 \leq \|x\|_1$$

where

$$\|x\|_2 = \left(\int_{-\infty}^{\infty} |f(x)|^2 d\mu(x) \right)^{0.5}$$

and

$$\|x\|_1 = \int_{-\infty}^{\infty} |f(x)| d\mu(x)$$

Proof: for finite dimensional normed space cases: we need to prove

$$\sqrt{x_1^2 + x_2^2 + \dots + x_n^2} \leq |x_1| + |x_2| + \dots + |x_n|$$

By squaring both sides, we can get the result. For continuous case, TODO

Theorem A.7.1. [\[7\]](#) For L^p normed space, we have

$$\|x\|_q \leq \|x\|_p$$

whenever $p \leq q$ where

$$\|x\|_q = \left(\int_{-\infty}^{\infty} |f(x)|^q d\mu(x) \right)^{1/q}$$

Proof: todo

Remark A.7.1. For complete description on L^p norms, see [7]

A.7.3 Young's inequality for product

Lemma A.7.3. If $a, b \geq 0$, and $p, q > 1, 1/p + 1/q = 1$, then

$$ab \leq a^p/p + b^q/q$$

Proof:

$$\log(a^p/p + b^q/q) \geq \log(a^p)/p + \log(b^q)/q = \log(a) + \log(b) = \log(ab)$$

where we use the fact of log is concave.

A.8 Useful properties of matrix

A.8.1 Matrix derivatives

Lemma A.8.1 (common matrix derivative in quadratic forms). [8] For $A \in \mathbb{R}^{m \times n}$, $x \in \mathbb{R}^n$, we have:

$$\begin{aligned}\frac{\partial a^T x}{\partial x} &= \frac{\partial x^T a}{\partial x} = a \\ \frac{\partial Ax}{\partial x} &= A \\ \frac{\partial BAx}{\partial x} &= BA \\ \frac{\partial x^T Ax}{\partial x} &= (A + A^T)x \\ \frac{\partial x^T Ax}{\partial x} &= 2A\end{aligned}$$

Lemma A.8.2. If $f(x) = g(Ax)$, $A \in \mathbb{R}^{n \times n}$, $x \in \mathbb{R}^n$ for some differentiable function $g(y)$, then

$$\nabla f = A^T \nabla g$$

In particular, $a \in \mathbb{R}^n$, then

$$\nabla a^T Ax = A^T x$$

A.8.2 Matrix inversion lemma

Lemma A.8.3 (matrix inversion lemma). [9, p. 120]

- $(E - FH^{-1}G)^{-1} = E^{-1} + E^{-1}F(H - GE^{-1}F)^{-1}GE^{-1}$
- $(E - FH^{-1}G)^{-1}FH^{-1} = E^{-1}F(H - GE^{-1}F)^{-1}$

Proof. (1) can be verified (2) expand the parenthesis using (1). □

Corollary A.8.0.1 (matrix inversion of rank one update). Let $H = I$, $F = \pm u \in \mathbb{R}^n$, and $G = \pm v \in \mathbb{R}^n$, we have

-

$$(E - uv)^{-1} = E^{-1} - \frac{E^{-1}uv^TE^{-1}}{1 + v^TE^{-1}u}$$

-

$$(E - uv)^{-1} = E^{-1} + \frac{E^{-1}uv^TE^{-1}}{1 - v^TE^{-1}u}$$

A.8.3 Block matrix

Lemma A.8.4. *Given an $(m \times p)$ matrix A with q row partions and s colun partitions and a $(p \times n)$ matrix B with s row partions and r colun partitions,*

$$A = \begin{pmatrix} A_{11} & A_{12} & \cdots & A_{1s} \\ A_{21} & A_{22} & \cdots & A_{2s} \\ \vdots & \vdots & \ddots & \vdots \\ A_{q1} & A_{q2} & \cdots & A_{qs} \end{pmatrix}, B = \begin{pmatrix} B_{11} & B_{12} & \cdots & B_{1r} \\ B_{21} & B_{22} & \cdots & B_{2r} \\ \vdots & \vdots & \ddots & \vdots \\ B_{s1} & B_{s2} & \cdots & B_{sr} \end{pmatrix},$$

then the matrix product

$$C = AB$$

can be formed blockwise, giving C as an $(m \times n)$ matrix with q row partitions and r column partitions. In particular,

$$C_{\alpha\beta} = \sum_{\gamma=1}^s A_{\alpha\gamma} B_{\gamma\beta}.$$

Lemma A.8.5 (sum of vector product to matrix product). *Consider column vectors $x_1, x_2, \dots, x_N \in \mathbb{R}^d$ and column vectors $y_1, y_2, \dots, y_N \in \mathbb{R}^d$. It follows that*

-

$$\sum_{i=1}^N x_i^T y_i = X_C^T Y_C,$$

where $X_C \in \mathbb{R}^{Nd}$ is a vector stacking all the x_1, \dots, x_N (similarly Y_C).

-

$$\sum_{i=1}^N x_i y_i^T = X_R^T Y_R,$$

where $X_R \in \mathbb{R}^{N \times d}$ is a matrix stacking all the x_1^T, \dots, x_N^T (similarly Y_R).

Lemma A.8.6 (block matrix inversion formula).

$$\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}^{-1} = \begin{pmatrix} A_{11}^{-1} + B_{12}B_{22}^{-1}B_{21} & -B_{12}B_{22}^{-1} \\ -B_{22}^{-1}B_{21} & B_{22}^{-1} \end{pmatrix} \\ = \begin{pmatrix} C_{11}^{-1} & -C_{11}^{-1}C_{12} \\ -C_{21}C_{11}^{-1} & A_{22}^{-1} + C_{21}C_{11}^{-1}C_{12} \end{pmatrix}$$

where

$$B_{22} = A_{22} - A_{21}A_{11}^{-1}A_{12}, B_{12} = A_{11}^{-1}A_{12}, B_{21} = A_{21}A_{11}^{-1}$$

and

$$C_{11} = A_{11} - A_{12}A_{22}^{-1}A_{21}, C_{12} = A_{12}A_{22}^{-1}, C_{21} = A_{22}^{-1}A_{21}$$

A.8.4 Matrix trace

Lemma A.8.7. • $\|A\|_F^2 = \text{Tr}(AA^T)$

Lemma A.8.8 (matrix trace).

- (linearity) $\text{Tr}(aA + bB) = a\text{Tr}(A) + b\text{Tr}(B)$
- (commutative) $\text{Tr}(AB) = \text{Tr}(BA)$
- (invariance under transposition) $\text{Tr}(A) = \text{Tr}(A^T)$
- (cyclic rule) $\text{Tr}(ABCD) = \text{Tr}(DABC) = \text{Tr}(CDAB) = \text{Tr}(BCDA)$ or $\text{Tr}(ABC) = \text{Tr}(CAB) = \text{Tr}(BCA)$

Proof. (1)(2)(3) can be proved directly from definition. (4) We can group three elements together and commute with the fourth. For example, we can group (ABC) together and commute with D to prove the first equality. \square

Corollary A.8.0.2.

- (invariance under similar transformation) $\text{Tr}(PAP^{-1}) = \text{Tr}(A)$
- $\text{Tr}(X^TY) = \text{Tr}(XY^T) = \text{Tr}(Y^TX) = \text{Tr}(YX^T)$

Proof. (1) Use cyclic rule. (2) Use invariance under transposition and commutative rule. \square

Lemma A.8.9 (common matrix derivative involving matrix trace). [8] Let $A, X, B \in \mathbb{R}^{m \times m}$. We have

$$\begin{aligned}\frac{\partial \text{Tr}(X)}{\partial X} &= I \\ \frac{\partial \text{Tr}(XA)}{\partial X} &= \frac{\partial \text{Tr}(AX)}{X} = A^T \\ \frac{\partial \text{Tr}(X^T A)}{\partial X} &= \frac{\partial \text{Tr}(AX^T)}{X} = A \\ \frac{\partial \text{Tr}(AXB)}{\partial X} &= \frac{\partial \text{Tr}(BAX)}{X} = A^T B^T \\ \frac{\partial \text{Tr}(AX^T B)}{\partial X} &= \frac{\partial \text{Tr}(BAX^T)}{X} = BA \\ \frac{\partial \text{Tr}(XX^T)}{\partial X} &= 2X \\ \frac{\partial \text{Tr}(XX)}{\partial X} &= 2X^T\end{aligned}$$

Additional, we have chain rule given by

$$\frac{\partial \text{Tr}(X^T A^T A X)}{\partial X} = \frac{\partial \text{Tr}(XX^T A^T A)}{\partial X} = \frac{\partial \text{Tr}(XX^T A^T A)}{\partial XX^T} \frac{\partial XX^T}{\partial X} = 2A^T A X.$$

A.8.5 Matrix elementary operator

Lemma A.8.10 (elementary operator matrix). Left multiply a matrix A by an elementary matrix is equivalent to performing row operation. The elementary operator matrix is given by

- (Interchange row i and j) For example, exchange row 2 and row 3:

$$R_1 = \begin{bmatrix} 1 & & & \\ & 0 & 1 & \\ & 1 & 0 & \\ & & & 1 \end{bmatrix}$$

- (Multiply row i by s) For example

$$R_2 = \begin{bmatrix} 1 & & & \\ & s & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

- (Add s times row i to row j) For example, add s times row 2 to row 3

$$R_3 = \begin{bmatrix} 1 & & & \\ & 1 & 0 & \\ & s & 1 & \\ & & & 1 \end{bmatrix}.$$

Note that $R_3 = R_1 R_2 \neq R_2 R_1$.

Lemma A.8.11 (elementary operator matrix). Right multiply a matrix A by an elementary matrix is equivalent to performing row operation. The elementary operator matrix is given by

- (Interchange column i and j) For example, exchange row 2 and row 3:

$$C_1 = \begin{bmatrix} 1 & & & \\ & 0 & 1 & \\ & 1 & 0 & \\ & & & 1 \end{bmatrix}$$

- (Multiply column i by s) For example

$$C_2 = \begin{bmatrix} 1 & & & \\ & s & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

- (Add s times column i to column j) For example, add s times column 2 to column 3

$$C_3 = \begin{bmatrix} 1 & & & \\ & 1 & s & \\ & & 1 & \\ & & & 1 \end{bmatrix}.$$

Note that $C_3 = C_1C_2 \neq C_2C_1$.

A.8.6 Matrix determinant

Lemma A.8.12 (properties of determinant).

- All elementary operator matrix has determinant 1.
- For matrix $A \in \mathbb{R}^{n \times n}$,

$$\det(kA) = k^n \det(A).$$
- $\det(AB) = \det(A)\det(B)$.
- All elementary operation on a matrix will not change its determinant.

A.9 Numerical integration

Definition A.9.1 (Newton-Cotes Formula). Suppose we want to evaluate $\int_a^b f(x)dx$. We can evaluate $f(x)$ at $n + 1$ equally spacing points $x_i = a + i(b - a)/n$, and then we approximate $f(x)$ by n degree of Lagrange polynomial and do the integral. Specifically, we have

$$\int_a^b f(x)dx \approx \int_a^b L(x)dx = \int_a^b \left(\sum_{i=0}^n f(x_i)l_i(x) \right) = \sum_{i=1}^n f(x_i) \int_a^b l_i(x)dx = \sum_{i=1}^n f(x_i)w_i$$

where L is the Lagrange polynomial of degree n , and $l_i(x), i = 0, \dots, n$ is the $(n+1)$ Lagrange polynomial basis, given as [Lemma 3.5.2](#).

Example A.9.1. Consider we use degree 1 Lagrange polynomial to approximate $f(x)$, then

$$L(x) = f(a)\frac{x-a}{b-a} + f(b)\frac{x-b}{a-b}$$

where $l_0(x) = \frac{x-a}{b-a}$ and $l_1(x) = \frac{x-b}{a-b}$. Then

$$w_0 = \int_a^b l_0(x)dx = \frac{1}{2}, w_1 = \int_a^b l_1(x)dx = \frac{1}{2}.$$

Table A.9.1: Closed Newton-Cotes Formula

Notation: $\int_a^b f(x)dx, f_i = f(x_i), x_i = a + i(b - a)/n$			
Degree	Name	Formula	Error term
1	Trapezoid rule	$\frac{b-a}{2}(f_0 + f_1)$	$-\frac{(b-a)^3}{12}f^{(2)}(\eta)$
2	Simpson's rule	$\frac{b-a}{6}(f_0 + 4f_1 + f_2)$	$-\frac{(b-a)^5}{2880}f^{(4)}(\eta)$
3	Simpson's 3/8 rule	$\frac{b-a}{8}(f_0 + 3f_1 + 3f_2 + f_3)$	$-\frac{(b-a)^3}{6480}f^{(5)}(\eta)$

Remark A.9.1 (Error analysis). For detailed error analysis, see [\[10, p. 252\]](#).

Remark A.9.2 (how to use). Usually, given the integral $\int_a^b f(x)dx$, we will first divide into smaller intervals and do the numerical integral on each interval and add them up. For example

$$\int_a^b f(x)dx = \int_a^{a+h} f(x)dx + \int_{a+h}^{a+2h} f(x)dx \dots + \int_{b-h}^b f(x)dx.$$

Lemma A.9.1 (Trapezoid rule and the error bound). *Given the integral $\int_a^b f(x)dx$, we have*

$$\int_a^b f(x)dx \approx \frac{b-a}{n} \left(\frac{f(a)}{2} + \sum_{k=1}^{n-1} f\left(a + k \frac{b-a}{n}\right) + \frac{f(b)}{2} \right)$$

where we divide $b-a$ into n subintervals, and we do the Trapezoid rule on each subinterval. And the error bound is

$$\frac{(b-a)^3}{12n^2} \max_{x \in [a,b]} f^{(2)}(x)$$

Proof. Note that on each subinterval, the error is $-\frac{(b-a/n)^3}{12} f^{(2)}(\eta)$. Sum up n terms, and we have upper bound

$$\frac{(b-a)^3}{12n^3} n \max_{x \in [a,b]} f^{(2)}(x)$$

□

Lemma A.9.2 (Midpoint rule and the error bound). *Given the integral $\int_a^b f(x)dx$, we have*

$$\int_a^b f(x)dx \approx \frac{b-a}{n} \left(\sum_{k=1}^n f\left(a + (k-0.5) \frac{b-a}{n}\right) \right)$$

where we divide $b-a$ into n subintervals, and we do the Trapezoid rule on each subinterval. And the error bound is

$$\frac{(b-a)^3}{n^2} K$$

A.9.1 Gaussian quadrature

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

which is exact when f is a polynomial.

Remark A.9.3. In Newton-Cotes formulas, we fix nodes and try to find suitable weights; in Gaussian quadrature, we use a weighted sum of function values at specified points within the domain of integration.

A.10 Vector calculus**Lemma A.10.1 (divergence theorem).**

$$\begin{aligned}\iiint_V (\nabla \cdot \mathbf{F}) dV &= \oiint_{S(V)} \mathbf{F} \cdot \hat{\mathbf{n}} dS \\ \iiint_V (\nabla \times \mathbf{F}) dV &= \oiint_{S(V)} \hat{\mathbf{n}} \times \mathbf{F} dS \\ \iiint_V (\nabla f) dV &= \oiint_{S(V)} \hat{\mathbf{n}} f dS\end{aligned}$$

Lemma A.10.2 (Laplacian product rule). *Given functions $u : \mathbb{R}^n \rightarrow \mathbb{R}$ and $v : \mathbb{R}^n \rightarrow \mathbb{R}$, we have*

$$\nabla^2(uv) = u\nabla^2 v + 2\nabla u \cdot \nabla v + v\nabla^2 u.$$

Proof. Directly use product rule. □

A.11 Numerical linear algebra computation complexity

Note A.11.1. [11, p. 606]

- For a $m \times n$ matrix multiplying a n dimensional vector, mn .
- For a $n \times n$ matrix multiplying a $n \times n$ matrix, n^3 (without optimization).
- For a $n \times n$ matrix, LU decomposition $2n^3/3$ (for symmetric matrix $n^3/3$).
- For a $m \times n$ matrix, Cholesky decomposition $4m^2n/3$ (for square matrix $4n^3/3$).
- For a $m \times n$ matrix, QR decomposition $4m^2n/3$ (for square matrix $4n^3/3$).

Note A.11.2 (solving triangular linear system). Let L be a $n \times n$ lower triangle matrix, the forward substitution algorithm for solving

$$Ly = d,$$

is given by

```
y(1) = d(1) / L(1,1);  
for i=2:n  
y(i) = (d(i) - L(i,1:i-1)* y(1:i-1))/L(i,i)  
end
```

This algorithm has complexity of $O(n^2)$.

Let U be a $n \times n$ upper triangle matrix, the backward substitution algorithm for solving

$$Ux = d,$$

is given by

```
x(n) = d(n)/U(n,n);  
for i = n - 1: -1 :1  
x(i) = (d(i) - U(i,i + 1:n)*x(i + 1:n) )/U(i,i)  
end
```

This algorithm has complexity of $O(n^2)$.

A.12 Distributions

Lemma A.12.1. [12, p. 579] Let K be an externally given parameter. We have

- $\int_{-\infty}^{\infty} \delta(x) dx = 1, x\delta(x) = 0, \int_{-\infty}^{\infty} f(x)\delta(x - K) dx = f(K).$
- $\delta(g(x)) = \sum_i \frac{\delta(x - x_i)}{|g'(x_i)|},$ where $x_i, i = 1, 2, \dots$ are the zeros of the function $g(x).$
- $\delta(\lambda x) = \frac{\delta(x)}{|\lambda|}, \delta(x - K) = \delta(K - x).$
- (step function definition)

$$H(x) \triangleq \frac{d}{dx} \max\{x, 0\}, H(x - A) \triangleq \frac{d}{dx} \max\{x - A, 0\}$$

- $H(x - K) + H(K - x) = 1.$
-

$$\frac{dH(x - K)}{dx} = \delta(x - K), \frac{dH(K - x)}{dx} = -\delta(x - K).$$

Proof. Use $H(x - K) + H(K - x) = 1$ to prove $\frac{dH(K - x)}{dx}$. □

A.13 Common integrals**Lemma A.13.1.**

•

$$\int_0^\infty e^{-ax^2} dx = \frac{1}{2} \sqrt{\frac{\pi}{a}}, \quad \int_{-\infty}^\infty e^{-ax^2} dx = \sqrt{\frac{\pi}{a}}$$

•

$$\int_0^\infty x e^{-ax^2} dx = \frac{1}{2a}, \quad \int_{-\infty}^\infty x e^{-ax^2} dx = 0$$

•

$$\int_0^\infty x^2 e^{-ax^2} dx = \frac{1}{4a} \sqrt{\frac{\pi}{a}}$$

•

$$\int_0^\infty x^n e^{-ax} dx = \frac{n!}{a^{n+1}}$$

•

$$\int_0^\infty x^m e^{-ax^2} dx = \frac{\Gamma((m+1)/2)}{2a^{(m+1)/2}}$$

A.14 Nonlinear root finding

A.14.1 Bisection method

Methodology A.14.1.

- **(Goal):** We want to find a $x^* \in \mathbb{R}$ such that $f(x^*) = 0$.
- **(Initial input):** Initial guess of l_0 and r_0 such that

$$f(l_0) < 0, f(r_0) > 0; \text{ or } f(l_0) > 0, f(r_0) < 0.$$

- **Repeat (i is the iteration index):**
 - Let $m = \frac{r_i + l_i}{2}$.
 - If $f(l_i)f(m) < 0$, then $l_{i+1} = l_i, r_{i+1} = m$.
 - If $f(l_i)f(m) > 0$ (then we must have $f(r_i)f(m) < 0$), then $l_{i+1} = m, r_{i+1} = r_i$.
 - If $f(l_i)f(m) = 0$, then m is the root.

A.14.2 Newton method

Methodology A.14.2.

- **(Goal):** We want to find a $x^* \in \mathbb{R}$ such that $f(x^*) = 0$. f is assumed to be differentiable.
- **(Initial input):** Initial guess of x_0 .
- **Repeat (i is the iteration index):**
 - Let $x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$.
 - If $f(x_{i+1}) = 0$, then x_{i+1} is the root.

A.14.3 Secant method

Methodology A.14.3 (Secant method).

- **(Goal):** We want to find a $x^* \in \mathbb{R}$ such that $f(x^*) = 0$. f is assumed to be differentiable.
- **(Initial input):** Initial guess of x_0, x_1 .
- **Repeat (i is the iteration index):**

– Let

$$x_{i+1} = x_i - f(x_i) \frac{x_i - x_{i-1}}{f(x_i) - f(x_{i-1})}$$

– If $f(x_{i+1}) = 0$, then x_{i+1} is the root.

Remark A.14.1 (derivation). Starting with initial guesses x_0, x_1 , we construct a first order approximation of $f(x)$ via

$$y = \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x - x_1) + f(x_1).$$

And we solve the root for the first-order approximation problem via

$$\frac{f(x_1) - f(x_0)}{x_1 - x_0}(x - x_1) + f(x_1) = 0 \implies x_2 = x_1 - f(x_1) \frac{x_1 - x_0}{f(x_1) - f(x_0)}.$$

Then we continue the process with x_1, x_2 .

Remark A.14.2 (convergence property).

- There is no guarantee on the global convergence to the root of f .
- Only when the initial values x_0 and x_1 are sufficiently close to the root, the iterates x_n will converge to the root.

A.15 Interpolation

A.15.1 cubic interpolation

Definition A.15.1 (the cubic spine line functional form). [13]

- Suppose x_1, x_2, \dots, x_n and y_1, y_2, \dots, y_n are known.
- A cubic spine line is given by

$$y(x) = a_i + b_i(x - x_i) + c_i(x - x_i)^2 + d_i(x - x_i)^3, x_i \leq x \leq x_{i+1}, i = 1, 2, \dots, n - 1.$$

- There are $4n - 4$ unknowns.
- Note that

$$\begin{aligned} y'(x) &= b_i + 2c_i(x - x_i) + 3d_i(x - x_i)^2, x_i < x < x_{i+1} \\ y''(x) &= 2c_i + 6d_i(x - x_i), x_i < x < x_{i+1} \\ y'''(x) &= 6d_i, x_i < x < x_{i+1} \end{aligned}$$

Definition A.15.2 (natural cubic spline condition). [13]

Let $h_i = x_{i+1} - x_i$

- (spline line passing data points): for $i = 1, 2, \dots, n - 1$, $a_i = y_i$; $a_{n-1} + b_{n-1}h_{n-1} + c_{n-1}h_{n-1}^2 + d_{n-1}h_{n-1}^3 = y_n$.
- (interpolating function is continuous); that is,

$$\lim_{x \rightarrow x_i^-} y(x) = \lim_{x \rightarrow x_i^+} y(x) \implies a_i + b_i h_i + c_i h_i^2 + d_i h_i^3 = a_{i+1}, \forall i = 1, 2, \dots, n - 2.$$

- (interpolating function is differentiable); note that the interpolating function is differentiable on interval, therefore we require that,

$$\lim_{x \rightarrow x_i^-} y'(x) = \lim_{x \rightarrow x_i^+} y'(x) \implies b_i + 2c_i h_i + 3d_i h_i^2 = b_{i+1}, \forall i = 1, 2, \dots, n - 2$$

:

- (interpolating function is twice differentiable and the second derivative at each endpoint is 0); that is,

$$\lim_{x \rightarrow x_i^-} y''(x) = \lim_{x \rightarrow x_i^+} y''(x) \implies c_i + 3d_i h_i = c_{i+1}, \forall i = 1, 2, \dots, n - 2,$$

and $y''(x_1) = y''(x_n) = 0$.

- these $4n - 4$ equations will solve the $4n - 4$ unknowns.

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