

Chapter 7

Nonlinear problems

7.1 Recalling few results of functions of several variables

7.1.1 Continuity

[7.1.1] Fact. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ and $a \in \mathbb{R}$. TFAE.

- i) The function f is continuous at a .
- ii) For each unit vector $u \in \mathbb{R}$, we have $\lim_{t \rightarrow 0^+} f(a + tu) = f(a)$.

[7.1.2] Fact. Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ and $a \in \mathbb{R}^n$. The following are not equivalent.

- i) The function f is continuous at a .
- ii) For each unit vector $u \in \mathbb{R}^n$, we have $\lim_{t \rightarrow 0^+} f(a + tu) = f(a)$.

To see this define

$$f(x, y) = \begin{cases} \frac{y^2}{x}, & \text{if } x \neq 0 \\ 0 & \text{else.} \end{cases}$$

Notice that for each unit vector $u \in \mathbb{R}^2$, we have $\lim_{t \rightarrow 0^+} f(0 + tu) = f(0)$. However, f is not continuous at 0 because if we approach 0 along the curve (imagine a sequence of points on this curve converging to $(0, 0)$) $y = \sqrt{x}$, the limit is 1, that is, $\lim_{\substack{x \rightarrow 0^+ \\ y=\sqrt{x}}} f(x, y) = 1$.

[7.1.3] Fact. Let $P(x)$ and $Q(x)$ be polynomials in $x = (x_1, \dots, x_n)$. Then the function $\frac{P(x)}{Q(x)}$ is continuous wherever it is defined on \mathbb{R}^n .

[7.1.4] Fact. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ such that both $g(y) = f(a, y)$, $h(x) = f(x, b)$ are continuous for each $a, b \in \mathbb{R}$. Even then, $f(x, y)$ may not be continuous.

To see this, consider the function $f(x, y) = \frac{xy}{x^2+y^2}$, $f(0, 0) = 0$. If $a \neq 0$ then $f(a, y) = \frac{ay}{a^2+y^2}$ is continuous on \mathbb{R} . If $a = 0$ then $f(a, y) = 0$ is continuous on \mathbb{R} . Similarly $f(x, b)$ is continuous for any fixed b . The function f is not continuous at $(0, 0)$: approaching $(0, 0)$ along the line $y = mx$, the limit is $\frac{m}{1+m^2}$, which changes with m .

7.1.2 Differentiability

[7.1.5] Discussion : coefficient matrix of linear terms in $f(a + h) - f(a)$.

- a) Consider a polynomial $f(x) = x_1^2 + x_1x_2$. Then

$$f(a + h) - f(a) = 2a_1h_1 + a_2h_1 + a_1h_2 + h_1^2 + h_1h_2.$$

Notice that, the linear terms can also be written as $[2a_1 + a_2 \quad a_1] \begin{bmatrix} h_1 \\ h_2 \end{bmatrix}$.

- b) So, ‘the coefficient matrix of linear terms in $f(a + h) - f(a)$ ’ is $A = [2a_1 + a_2 \quad a_1]$.

- c) Thus, for $f(x) = x_1^2 - x_1x_2 + x_3^2$ defined on \mathbb{R}^3 and $a = (1, 2, 3)$, the coefficient matrix of the linear terms in h in the expression $f(a + h) - f(a)$ is $[2a_1 - a_2 \quad -a_1 \quad 2a_3]$.

To get this quickly, try to imagine the terms that could give you $(\dots)h_1$.

[7.1.6] Fact. Let $f(x)$ be a real polynomial in $x = (x_1, \dots, x_n)$ and $a \in \mathbb{R}^n$. Let A be the coefficient of linear terms in h of $f(a + h) - f(a)$. Then $\lim_{\|h\| \rightarrow 0} \frac{|f(a+h) - f(a) - Ah|}{\|h\|} = 0$.

Proof. Note that $f(a + h) - f(a) - Ah$ is a polynomial with terms of degree two or more in h . So $\lim_{\|h\| \rightarrow 0} \frac{|f(a+h) - f(a) - Ah|}{\|h\|} = 0$. Note that $\lim_{\|h\| \rightarrow 0} \frac{|h_1h_2|}{\|h\|} = \lim_{\|h\| \rightarrow 0} \frac{|h_1|}{\|h\|}|h_2| = 0$. ■

[7.1.7] Discussion : differentiability.

- a) Let $E \subseteq \mathbb{R}^n$ be open, $f : E \rightarrow \mathbb{R}$, and $a \in E$. Suppose that there is a matrix $A_{1 \times n}$ such that

$$\lim_{\|h\| \rightarrow 0} \frac{|f(a + h) - f(a) - Ah|}{\|h\|} = 0.$$

Then we say f is DIFFERENTIABLE at a and write $f'(a) = A$. We sometimes call A , the TOTAL DERIVATIVE of f at a . The GRADIENT $\nabla f(a)$ of f at a is the vector $f'(a)^t$.

- b) Thus for $f(x, y, z) = x^2 + y^2 + xyz$, we have $f'(x, y, z) = [2x + yz \quad 2y + xz \quad xy]$.

[7.1.8] Discussion : directional derivatives.

- a) On your way to school, the height of the road increases and the road goes as if it is the line $f(x) = x$. What is the slope of the road we are facing?

- b) What is the slope we will face on our way back?

- c) So, the slope changes according to which direction we are facing.
- d) Let $E \subseteq \mathbb{R}^n$ be open, $f : E \rightarrow \mathbb{R}$, $a \in E$ and $0 \neq u \in \mathbb{R}^n$. The DIRECTIONAL DERIVATIVE $D_u f(a)$ of f at a in the direction of u is defined as

$$D_u f(a) := \lim_{t \rightarrow 0} \frac{f(a + tu) - f(a)}{t},$$

provided the limit exists. It means the ‘instantaneous rate of change of f where unit step means u ’.

- e) Thus, for $f(x, y, z) = x^2 + y^2 + xyz$, $a = (1, 2, 3)$, $u = (1, 0, 1)$ and $v = (0, 1, 1)$, we have

$$D_u f(a) = \lim_{t \rightarrow 0} \frac{(1+t)^2 + 2^2 + (1+t)2(3+t) - 1^2 - 2^2 - 1 \cdot 2 \cdot 3}{t} = 10$$

and $D_v f(a) = 9$.

- f) When $u = e_i$ the directional derivative $D_u f(a)$ is called the PARTIAL DERIVATIVE $D_i f(a)$ of f with respect to the i th coordinate.

[7.1.9] Fact. Suppose that $D_u f(a) = \beta$ and $\alpha \neq 0$. Then $D_{\alpha u} f(a)$ exists and $D_{\alpha u} f(a) = \alpha D_u f(a)$.

Proof. We have $D_{\alpha u} f(a) = \lim_{t \rightarrow 0} \frac{f(a + t\alpha u) - f(a)}{t} = \alpha \lim_{t \rightarrow 0} \frac{f(a + t\alpha u) - f(a)}{\alpha t} = \alpha \beta = \alpha D_u f(a)$. ■

[7.1.10] Fact. Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ such that $D_u f(a)$ exists for each $u \neq 0$. Even then f need not be continuous at a .

For example, take $f(x, y) = \frac{y^2}{x}$ if $x \neq 0$ and 0, otherwise. Notice that $D_1 f(0) = \lim_{t \rightarrow 0} \frac{f(t, 0) - f(0, 0)}{t} = 0$ and $D_2 f(0) = \lim_{t \rightarrow 0} \frac{f(0, t) - f(0, 0)}{t} = 0$. Take a unit vector $u = \left[\frac{1}{\sqrt{m^2+1}} \quad \frac{m}{\sqrt{m^2+1}} \right]^t$ on the line $y = mx$, we have

$$D_u f(0) = \lim_{t \rightarrow 0} \frac{f(tu_1, tu_2) - f(0)}{t} = \frac{m^2}{\sqrt{m^2+1}}.$$

Thus the directional derivatives exist in all directions. We already know that f is not continuous at 0. ■

[7.1.11] Fact. Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be continuous at a . Then $D_u f(a)$ may not exist even for a single $u \neq 0$. For example, take $f(x) = \|x\|$ on \mathbb{R}^n and check at 0.

[7.1.12] Fact. Let $E \subseteq \mathbb{R}^n$ be open, $f : E \rightarrow \mathbb{R}$, $a \in E$, and $0 \neq u \in \mathbb{R}^n$. Suppose that f is differentiable at a with $f'(a) = A = [d_1 \dots d_n]$. Then f is continuous at a and $D_u f(a)$ exists with $D_u f(a) = \langle \nabla f(a), u \rangle$. In particular, $D_i f(a) = d_i$.

Proof. To prove continuity note that by definition $\lim_{\|h\| \rightarrow 0} \frac{|f(a+h) - f(a) - Ah|}{\|h\|} = 0$. So $\lim_{\|h\| \rightarrow 0} |f(a+h) - f(a) - Ah| = 0$ and $\lim_{\|h\| \rightarrow 0} f(a+h) = f(a)$. So f is continuous.

To prove the next assertion note that as $\lim_{\|h\| \rightarrow 0} \frac{|f(a+h) - f(a) - Ah|}{\|h\|} = 0$, in particular, taking $h = tu$, we have $\lim_{t \rightarrow 0} \left| \frac{f(a+tu) - f(a) - tAu}{t} \right| = 0$. That is, $\lim_{t \rightarrow 0} \left| \frac{f(a+tu) - f(a)}{t} - Au \right| = 0$. That is, $\lim_{t \rightarrow 0} \frac{f(a+tu) - f(a)}{t} = Au$. That is, $D_u f(a) = Au = \langle \nabla f(a), u \rangle$. The proof is complete. ■

To verify whether a function is differentiable at a point

1. Find all $D_i f(a)$. If some $D_i f(a)$ does not exist, then conclude that f is not differentiable at a .
2. Form the matrix $A = [D_1 f(a) \cdots D_n f(a)]$. Substitute A in the limit definition of the derivative and check if the limit is 0.
3. Conclude 'yes', if the limit is 0, otherwise conclude 'no'.

[7.1.13] Example. Take $f(x, y) = \begin{cases} \frac{x^2 y}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0). \end{cases}$ Check whether it is differentiable at 0.

Answer. We have $D_1 f(0) = 0$ and $D_2 f(0) = 0$. But $\lim_{h \rightarrow 0} \left| \frac{f(h) - [0 \ 0]^T h}{\|h\|} \right|$ does not exist. So it is not differentiable at 0. \blacksquare

[7.1.14] Fact : a sufficient condition for differentiability. Let $E \subseteq \mathbb{R}^n$ be open and $f : E \rightarrow \mathbb{R}$. If $D_i f$ are continuous on E , then f is differentiable on E .

This condition is not necessary for differentiability. For example, take $f : (-1, 1) \rightarrow \mathbb{R}$ defined as

$$f(x) = \begin{cases} x^2 \sin(1/x) & \text{if } x \neq 0 \\ 0 & \text{if } x = 0. \end{cases}$$

Then $f'(0) = 0$ and $f'(x) = 2x \sin(1/x) - \cos(1/x)$ for $x \neq 0$. But f' is not continuous at 0.

[7.1.15] Exercise(E).

- Let $f(x, y) = \min\{x, y\}$. Check the directional derivatives and differentiability at 0.
- Let $f(x, y) = \min\{x^2, y^2\}$. Check the directional derivatives and differentiability at 0, $(1, 1)$ and $(1, 2)$.

[7.1.16] Discussion : higher order derivatives.

- Let $E \subseteq \mathbb{R}^n$ be open and $f : E \rightarrow \mathbb{R}$. Then f is said to be CONTINUOUSLY DIFFERENTIABLE on E , denoted $f \in \mathcal{C}^1(E)$, if $D_i f$ are continuous on E .
- Let $E \subseteq \mathbb{R}^n$ be open and $f : E \rightarrow \mathbb{R}$. Then the SECOND ORDER PARTIAL DERIVATIVE $D_{ij} f$ is defined as $D_{ij} f = D_i(D_j f)$, if it exists.
- It can happen that $D_{ij} f$ and $D_{ji} f$ both exist and unequal at a point.

For example, consider

$$f(x, y) = \begin{cases} \frac{xy(x^2 - y^2)}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0). \end{cases}$$

- But, it is known from calculus that, if $D_{ij} f, D_{ji} f$ are continuous on E (open), then $D_{ij} f = D_{ji} f$ on E .

- e) Let $E \subseteq \mathbb{R}^n$ be open and $f : E \rightarrow \mathbb{R}^m$. Then f is said to be **TWICE CONTINUOUSLY DIFFERENTIABLE** on E , denoted $f \in \mathcal{C}^2(E)$, if all $D_{ij}f$ are continuous on E .
- f) All polynomials in x_1, \dots, x_n infinitely differentiable functions.
- g) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a \mathcal{C}^2 function. The **HESSIAN** $H_f(a)$ or $H(a)$ of f at a is the matrix (it is a real symmetric matrix as the function is \mathcal{C}^2)

$$H(a) = \begin{bmatrix} D_{11}f & D_{12}f & \cdots & D_{1n}f \\ D_{21}f & D_{22}f & \cdots & D_{2n}f \\ \vdots & \vdots & \ddots & \vdots \\ D_{n1}f & D_{n2}f & \cdots & D_{nn}f \end{bmatrix}(a).$$

- h) Thus the Hessian of $e^{x+y} - xyz$ at $(1, 1, 0)$ is $\begin{bmatrix} e^2 & e^2 & -1 \\ e^2 & e^2 & -1 \\ -1 & -1 & 0 \end{bmatrix}$.

[7.1.17] **Class workout.** Let $f : \mathbb{R}^3 \rightarrow \mathbb{R}$ be a \mathcal{C}^2 function and $a \in \mathbb{R}^3$.

- a) Is it true that $\sum_{i=1}^3 D_i f(a)x_i = \langle \nabla f(a), x \rangle$?
- b) Is it true that $\sum_{i,j=1}^3 D_{ij}f(a)x_i x_j = x^t H(a)x$?

7.1.3 Taylor's theorems

[7.1.18] **Fact.** (Rolle's theorem : single variable.) Let f be continuous on $[a, b]$ and differentiable on (a, b) with $f(a) = f(b)$. Then $\exists c \in (a, b)$ where $f'(c) = 0$.

[7.1.19] **Fact.** (Generalized mean value theorem.) Let f, g be continuous on $[a, b]$ and differentiable on (a, b) . Then $\exists c \in (a, b)$ where $f'(c)[g(b) - g(a)] = g'(c)[f(b) - f(a)]$.

Proof. (Self) Consider $H(x) = f(x)[g(b) - g(a)] - g(x)[f(b) - f(a)]$. Apply Rolle's theorem. ■

[7.1.20] **Fact.** (Taylor's theorem in one variable.) Let $E \subseteq \mathbb{R}$ be open, $f : E \rightarrow \mathbb{R}$ be in $C^n(E)$ and $[a, a+x] \subseteq E$. Then $\exists t \in (a, a+x)$ such that

$$f(a+x) = f(a) + f'(a)x + \frac{1}{2!}f''(a)x^2 + \cdots + \frac{1}{(n-1)!}f^{(n-1)}(a)x^{n-1} + \frac{1}{n!}f^{(n)}(t)x^n.$$

Proof. (Self) For $r \in [0, x]$, consider

$$F(r) = f(a+r) + f'(a+r)(x-r) + \cdots + \frac{1}{(n-1)!}f^{(n-1)}(a+r)(x-r)^{n-1}, \quad G(r) = (x-r)^n.$$

Then F, G are continuous on $[0, x]$ and differentiable on $(0, x)$. Apply generalized mean value theorem: $\exists c \in (0, x)$ such that $F'(c)[G(x) - G(0)] = G'(c)[F(x) - F(0)]$. So

$$\left(f'(a+c) + f''(a+c)(x-c) - f'(a+c) + f^{(3)}(a+c)\frac{(x-c)^2}{2!} - f''(a+c)(x-c) + \cdots + f^{(n)}(a+c)\frac{(x-c)^{n-1}}{(n-1)!}\right. \\ \left.- f^{(n-1)}(a+c)\frac{(x-c)^{n-2}}{(n-2)!}\right)[-x^n] = -n(x-c)^{n-1} \left[f(a+x) - f(a) - f'(a)x - \cdots - f^{(n-1)}(a)\frac{x^{n-1}}{(n-1)!}\right].$$

So

$$f^{(n)}(a+c)\frac{(x-c)^{n-1}}{(n-1)!}x^n = n(x-c)^{n-1} \left[f(a+x) - f(a) - f'(a)x - \cdots - f^{(n-1)}(a)\frac{x^{n-1}}{(n-1)!}\right].$$

That is, $f^{(n)}(a+c)\frac{x^n}{n!} = [f(a+x) - f(a) - f'(a)x - \cdots - f^{(n-1)}(a)\frac{x^{n-1}}{(n-1)!}]$. So, $\exists t := a+c \in (a, a+x)$ such that

$$f(a+x) = f(a) + f'(a)x + \frac{1}{2!}f''(a)x^2 + \cdots + \frac{1}{(n-1)!}f^{(n-1)}(a)x^{n-1} + \frac{1}{n!}f^{(n)}(t)x^n.$$

[7.1.21] Fact (Taylor's theorem in many variables) Let $E \subseteq \mathbb{R}^n$ be open, $f : E \rightarrow \mathbb{R}$ be in $C^m(E)$ and $[a, a+x] \subseteq E$. Then $\exists t \in (0, 1)$ such that

$$f(a+x) = f(a) + \sum_i D_i f(a) x_i + \frac{1}{2!} \sum_{i,j} D_{ij} f(a) x_i x_j + \cdots + \frac{1}{(m-1)!} \sum_{i_1, \dots, i_{m-1}} D_{i_1 \dots i_{m-1}} f(a) x_{i_1} \cdots x_{i_{m-1}} \\ + \frac{1}{m!} \sum_{i_1, \dots, i_m} D_{i_1 \dots i_m} f(a+tx) x_{i_1} \cdots x_{i_m}.$$

Proof. (Self [Rudin-p243]) Define $p(t) = a + tx$, for $0 \leq t \leq 1$. Define $h(t) = f(p(t))$. For any $t \in (0, 1)$, by chain rule,

$$h'(t) = f'(p(t))p'(t) = \sum_{i=1}^n D_i f(p(t))x_i, \quad h''(t) = \sum_{i,j} x_i x_j D_{ij} f(p(t)),$$

and so on. By Taylor's theorem in one variable, $\exists t \in (0, 1)$ such that $h(1) = \sum_{k=0}^{m-1} \frac{h^{(k)}(0)}{k!} + \frac{h^{(m)}(t)}{(m)!}$.

[7.1.22] Discussion. (Another look of Taylor's theorem.)

- a) Assume that $a = (a_1, \dots, a_n)^t$. Take $\alpha = (\alpha_1, \dots, \alpha_n)$ where $\alpha_i \geq 0$ are integers with $|\alpha| := \alpha_1 + \cdots + \alpha_n = k$. That is, α is a solution to the equation $z_1 + \cdots + z_n = k$ in nonnegative integers. There are $\binom{n+k-1}{k}$ many such solutions.
- b) Use the notations $D^\alpha := D_1^{\alpha_1} D_2^{\alpha_2} \cdots D_n^{\alpha_n}$, $x^\alpha := x_1^{\alpha_1} x_2^{\alpha_2} \cdots x_n^{\alpha_n}$, $\alpha! := \alpha_1! \cdots \alpha_n!$.
- d) Then the expression in Taylor's theorem will look like

$$f(a+x) = f(a) + \sum_{|\alpha|=1} D^\alpha f(a) \frac{x^\alpha}{\alpha!} + \cdots + \sum_{|\alpha|=m-1} D^\alpha f(a) \frac{x^\alpha}{\alpha!} + \sum_{|\alpha|=m} D^\alpha f(a+tx) \frac{x^\alpha}{\alpha!}.$$

- a) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be infinitely differentiable. Then the TAYLOR SERIES $T_f(\mathbf{x}; \mathbf{a})$ of f about the point \mathbf{a} is defined as

$$f(\mathbf{a}) + \sum_{|\alpha|=1} D^\alpha f(\mathbf{a}) \frac{(\mathbf{x}-\mathbf{a})^\alpha}{\alpha!} + \sum_{|\alpha|=2} D^\alpha f(\mathbf{a}) \frac{(\mathbf{x}-\mathbf{a})^\alpha}{\alpha!} + \sum_{|\alpha|=3} D^\alpha f(\mathbf{a}) \frac{(\mathbf{x}-\mathbf{a})^\alpha}{\alpha!} + \dots$$

- b) (Example.) Take $f(y, z) = y^2z^4 + yz^3 - 5yz + 6$ and $\mathbf{a} = (1, 2)$. Find the coefficient of $(y-1)^2(z-2)^2$ in $T_f((y, z); \mathbf{a})$ in two different ways.

Answer. Use $\mathbf{x} = (y, z)$. We want $(\mathbf{x}-\mathbf{a})^\alpha = (y-1)^2(z-2)^2$, that is, $\alpha = (2, 2)$. Hence, the coefficient is

$$\frac{1}{\alpha!} D_{1,1,2,2} f(\mathbf{a}) = \frac{1}{2!2!} (2.4.3.2^2) = 24.$$

o Alternately, note that

$$f(y, z) = (y-1+1)^2(z-2+2)^4 - (y-1+1)(z-2+2)^3 - 5(y-1+1)(z-2+2) + 6.$$

The coefficient for $(y-1)^2(z-2)^2$ can only come from the first term. When expanded using binomial expansion, it will look like $(y-1)^2 \binom{4}{2} (z-2)^2 2^2$. So the required coefficient is 24.

7.2 Positive definite matrices

[7.2.1] **Definition.** Let $A \in M_n(\mathbb{C})$.

- a) It is called POSITIVE DEFINITE(pd) if $x^*Ax > 0$ holds $\forall x \in \mathbb{C}^n \setminus \{0\}$.
 b) It is called POSITIVE SEMIDEFINITE(psd) if $x^*Ax \geq 0$ holds $\forall x \in \mathbb{C}^n$.

[7.2.2] **Facts and examples.**

- a) Let $A \in M_n(\mathbb{C})$ be pd. Then $a_{ii} = e_i^T A e_i > 0$.

- a') Let $A \in M_n(\mathbb{C})$ be psd. Then $a_{ii} = e_i^T A e_i \geq 0$.

- b) Take any matrix $A \in M_n(\mathbb{C})$. Then A^*A and AA^* are psd. If A is nonsingular, then they are pd.

Proof. (Self) For each x , we have $x^*A^*Ax = (Ax)^*Ax = \|Ax\|^2 \geq 0$. If $x \neq 0$ and A is nonsingular, then $Ax \neq 0$ and so $\|Ax\|^2 > 0$. ■

- c) The matrix I is pd. The zero matrix is a psd matrix.

- d) A pd matrix is by definition a psd matrix.

- e) A singular matrix cannot be pd.

[7.2.3] **Fact. (Psd implies Hermitian)** Let $A \in M_n(\mathbb{C})$ be a psd matrix. Then $A^* = A$.

Proof. (Self) Fix i, j , $i \neq j$. Let $v = e_i + e_j$. As $v^*Av \geq 0$, we have $a_{ii} + a_{ij} + a_{ji} + a_{jj} \geq 0$. But we already know that $a_{ii} + a_{jj} \geq 0$. So, we get $a_{ij} + a_{ji} \in \mathbb{R}$. That is, $\text{Im}(a_{ij}) = -\text{Im}(a_{ji})$.

Now take $v = e_i + ie_j$. As $v^*Av \geq 0$, we have $a_{ii} + ia_{ij} - ia_{ji} + a_{jj} \geq 0$. But we already know that $a_{ii} + a_{jj} \geq 0$. So, we get $a_{ij} - a_{ji} \in i\mathbb{R}$. That is, $\operatorname{Re}(a_{ij}) = \operatorname{Re}(a_{ji})$.

So $\overline{a_{ij}} = a_{ji}$.

[7.2.4] **Lemma.** Let $A \in M_n(\mathbb{R})$.

- a) Then A is pd iff $A^t = A$ and $x^t Ax > 0$ holds $\forall x \in \mathbb{R}^n$, $x \neq 0$.
- b) Then A is psd iff $A^t = A$ and $x^t Ax \geq 0$ holds $\forall x \in \mathbb{R}^n$.

Proof. (Self) a) Let A be a real pd matrix. As A is pd, it is Hermitian, and in our case it is symmetric. and $x^t Ax > 0$ for all real $x \neq 0$.

Conversely, suppose that A is real symmetric and $x^t Ax > 0$ for all real $x \neq 0$. Note that, for a real symmetric matrix A , we have $x^t Ay = y^t Ax$. Hence

$$(x + iy)^* A (x + iy) = x^t Ax + y^t Ay + i(x^t Ay - y^t Ax) = x^t Ax + y^t Ay > 0,$$

if $x + iy \neq 0$. The proof for b) is similar.

[7.2.5] **Theorem.** (Equivalent conditions for pd/psd.) Let $A \in M_n(\mathbb{C})$. TFAE.

- a) A is pd (resp. psd).
- b) $A^* = A$ and the eigenvalues of A are positive (resp. nonnegative).
- c) $A = B^*B$ for some nonsingular (resp. not necessarily nonsingular) matrix B .

Proof. (Self) a) \Rightarrow b). Let A be pd. We already know that $A^* = A$. Let (λ, x) be an eigenpair. Then $x^*Ax = x^*x\lambda \Rightarrow \lambda = \frac{x^*Ax}{x^*x} > 0$, as A is pd.

b) \Rightarrow c). Suppose that $A^* = A$ and all eigenvalues of A are positive. Then, by spectral theorem (for Hermitian matrices), there exists a unitary matrix U and a diagonal matrix $\Lambda = \operatorname{diag}(\lambda_1, \dots, \lambda_n)$, $\lambda_i > 0$ such that $A = U^*\Lambda U$. So

$$A = U^*\Lambda^{1/2}\Lambda^{1/2}U = U^*\Lambda^{1/2}UU^*\Lambda^{1/2}U = B^*B.$$

Notice that B is nonsingular as Λ has positive diagonal entries.

c) \Rightarrow a). Already done.

The proof for the psd part is similar.

We will need the following notations.

[7.2.6] **Definition.**

- a) Let $A \in M_n(\mathbb{C})$ and $S \subseteq [n]$. By $A(S, S)$, we denote the submatrix of A whose entries are indexed by S . This is called the PRINCIPAL SUBMATRIX of A indexed by S .

- b) For $A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 2 & 4 & 3 \\ 0 & 2 & 3 & 1 \\ 9 & 8 & 7 & 6 \end{bmatrix}$ and $S = \{1, 4\}$, we have $A(S, S) = \begin{bmatrix} 1 & 4 \\ 9 & 6 \end{bmatrix}$.

- c) A LEADING PRINCIPAL SUBMATRIX is one for which the subset S is $\{1, 2, \dots, k\}$ for some k .
- d) For the matrix A in b), and $S = \{1, 2\}$, we have $A(S, S) = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$. It is a leading principal submatrix.
- e) The matrix A in b), has only four leading principal submatrices.
- f) The determinant of a square submatrix is called a MINOR. The determinant of a leading principal submatrix is called a ‘leading principal minor’.
- g) Sometimes we use $A(S|S)$ to mean the matrix $A(S^c, S^c)$.

We will need the following two results from advanced linear algebra.

[7.2.7] Theorem. (Cauchy interlacing theorem.) Let A and $B = \begin{bmatrix} A & y \\ y^* & b \end{bmatrix}$ be Hermitian. Let $\lambda_1 \leq \dots \leq \lambda_n$ be the eigenvalues of A and $\mu_1 \leq \dots \leq \mu_{n+1}$ be the eigenvalues of B . Then

$$\mu_1 \leq \lambda_1 \leq \mu_2 \leq \lambda_2 \leq \dots \leq \lambda_n \leq \mu_{n+1}.$$

That is, the eigenvalues of A and B interlace.

[7.2.8] Theorem. The eigenvalues of a matrix A are continuous functions of its entries.

Now we can give two useful conditions to test whether a matrix is pd/psd.

[7.2.9] Theorem. (Test for pd/psd.) Let $A \in M_n(\mathbb{C})$.

- a) Then A is pd iff $A^* = A$ and the leading principal minors of A are positive.
- b) Then A is psd iff $A^* = A$ and all principal minors are nonnegative.

Proof. (Self) a) Suppose that A is pd. Then $A^* = A$ and the eigenvalues $\lambda_i > 0$. By Cauchy interlacing theorem, the eigenvalues of the leading principal matrices are > 0 . Hence their determinant is positive.

Conversely, suppose that $A^* = A$ and the leading principal minors are positive. By use of interlacing, we see that eigenvalues of A are positive. So A is pd.

b) If A is psd, then showing all principal minors ≥ 0 is similar to that of a). To prove the converse, let $A^* = A$ and assume that all principal minors are ≥ 0 . Note that

$$|A + \epsilon e_1 e_1^t| = (a_{11} + \epsilon)|A(1|1)| - a_{12}|A(1|2)| + a_{13}|A(1|3)| - \dots = |A| + \epsilon|A(1|1)| \geq |A|,$$

so that adding $\epsilon > 0$ to a diagonal entry does not decrease the determinant. Notice that we get $|A + \epsilon e_1 e_1^t| > |A|$, if $|A(1|1)| > 0$ and $\epsilon > 0$.

Next we show that $|A + \epsilon I| > 0$ whenever A has all minors nonnegative. We proceed by induction. Assume that it is true for $n - 1$. Consider the case for n and let B be obtained by adding ϵ to all diagonal entries of A except the first. Notice that $A(1|1)$ is Hermitian and it has all principal minor ≥ 0 . By induction hypothesis, we have that $|B(1|1)| > 0$. Hence by earlier argument $|B| > 0$.

It now follows that $A + \epsilon I$ is pd. This is true for each $\epsilon > 0$. As eigenvalues are continuous functions of the entries, it follows that eigenvalues of A are nonnegative. Hence A is psd.

[7.2.10] NoPen.

- a) If we say ‘a matrix $A \in M_n(\mathbb{C})$ is psd iff $A^* = A$ and all leading principal minors are nonnegative’, what would go wrong?
- b) Let $A \in M_n(\mathbb{C})$ be pd. Should every (nonempty) principal submatrix of A be pd?
- c) Let $A \in M_n(\mathbb{C})$ have positive eigenvalues. Must it be similar to a positive definite matrix?
- d) If A is pd and S is a nonsingular matrix, is S^*AS necessarily pd?
- e) How do I create a pd matrix $A \in M_{20}(\mathbb{R})$ which does not have a zero entry?
- f) Let $f \in C^2(E)$, $E \subseteq \mathbb{R}^n$ be open and $a \in E$. Suppose that $B_\delta(a) \subseteq E$ and for each $x \in B_\delta(a)$, $x \neq a$, we see that $H(x)$ is a pd matrix. Can we conclude that $H(a)$ is a psd matrix?

[7.2.11] Exercise(E+) Let $A \in M_n(\mathbb{C})$ be a positive definite matrix and $\lambda > 0$ be the smallest eigenvalue of A . Show that, for each $x \in \mathbb{C}^n$, we have $x^*Ax \geq \lambda x^*x$.

7.3 Unconstrained optimization

For now, let us consider the general problem of the form

$$(P1) \quad \begin{array}{ll} \min & f(x) \\ \text{s.t.} & x \in T \subseteq \mathbb{R}^n. \end{array}$$

[7.3.1] Definition. Consider (P1). Let $T \subseteq \mathbb{R}^n$ be nonempty and $a \in T$. A FEASIBLE DIRECTION at a is a vector d (allow $d = 0$) such that we can travel from a in that direction some positive multiple amount and while staying inside T . That is, $\exists \delta > 0$ such that $a + \theta d \in T$, $\forall \theta \in (0, \delta)$.

◦ By $D(a)$, we denote the set of all feasible directions available at a . (Remember that, 0 is allowed as a feasible direction.) We shall use $\overline{D}(a)$ to denote the closure of $D(a)$.

◦ Feasible directions at a point a only depend the structure of the feasible region near that point.

[7.3.2] Example. Consider the shaded region in the following figure.