

MATH 2

WEEKS 1-4

① Matrix multiplication

$$A \times B = C \quad ; \quad C[i,j] = \sum_{k=1}^n A[i,k] \times B[k,j]$$

② Determinant

$$\text{For } A_{2 \times 2} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc$$

$$\text{For } A_{3 \times 3} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = a_{11} \cdot \det \begin{bmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{bmatrix} - a_{12} \cdot \det \begin{bmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{bmatrix} + a_{13} \cdot \det \begin{bmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix}$$

$$\textcircled{3} \det(A \cdot B) = \det(A) \cdot \det(B)$$

$$\textcircled{4} \det(A^n) = \det(A)^n \quad ; \quad \det(A^{-1}) = \det(A)^{-1}$$

$$\textcircled{5} \det(A) = \text{product of diagonals if } A \text{ is an upper/lower triangle matrix}$$

$$\textcircled{6} \det(A) = \det(A^T)$$

⑦ Minor $\rightarrow (i,j)$ -th minor M_{ij} is the determinant of the submatrix after deleting i -th row and j -th column.
Cofactors $\rightarrow (i,j)$ -th cofactor C_{ij} is $(-1)^{i+j} \cdot M_{ij}$

$$\textcircled{8} \det(A_{n \times n}) = \sum_{j=1}^n a_{1j} C_{1j}$$

$$\textcircled{9} \det(t \cdot A_{n \times n}) = (t)^n \cdot \det(A)$$

⑩ Inverse matrix . If $\det(A) \neq 0$, then A is invertible

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

$$\text{For } A \text{ bigger than } 2 \times 2 \rightarrow A^{-1} = \frac{\text{adj}(A)}{\det(A)} \quad , \text{ where } \text{adj}(A) = C^T \rightarrow \text{transpose of cofactor matrix}$$

$$\text{For } A_{n \times n}, \det(\text{adj}(A)) = (\det(A))^{n-1}$$

⑪ Homogeneous System of Linear Equations

$$Ax = b, \text{ where } b = 0$$

\rightarrow has a unique solution 0 if A is invertible

⑫ Elementary Row Operations

Type	Action	Example and notation	Effect on determinant
1	Interchange two rows	$\begin{bmatrix} 3 & 2 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 7 & 1 & 1 \end{bmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 3 & 2 & 1 & 1 \\ 0 & 7 & 1 & 1 \end{bmatrix}$	$\det(A) = -\det(B)$
2	Scalar multiplication of a row by a constant t	$\begin{bmatrix} 3 & 2 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 7 & 1 & 1 \end{bmatrix} \xrightarrow{R_1/3} \begin{bmatrix} 1 & 2/3 & 1/3 & 1/3 \\ 1 & 1 & 0 & 0 \\ 0 & 7 & 1 & 1 \end{bmatrix}$	$\det(A) = t \cdot \det(B)$
3	Adding multiples of a row to another row	$\begin{bmatrix} 3 & 2 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 7 & 1 & 1 \end{bmatrix} \xrightarrow{R_1 - 3R_3} \begin{bmatrix} 3 & -19 & -2 & -2 \\ 1 & 1 & 0 & 0 \\ 0 & 7 & 1 & 1 \end{bmatrix}$	$\det(A) = \det(B)$

⑬ dependent variable \rightarrow if the column corresponding to that variable has the leading entry of some row.

⑭ Properties of vectors:

$$\text{i) } v + w = w + v$$

$$\text{v) } 1 \cdot (v) = v$$

$$\text{ii) } (v+w) + u = v + (w+u)$$

$$\text{vi) } (ab)v = a(bv)$$

$$\text{iii) } v + 0 = 0 + v = v$$

$$\text{vii) } a(v+w) = av + aw$$

$$\text{iv) } v + (-v) = 0$$

$$\text{viii) } (a+b)v = av + bv$$

- ⑮ A set of vectors is linearly independent iff the matrix V obtained by arranging the vectors in columns is invertible, i.e., $\det(V) \neq 0$.
- ⑯ If a set of vectors is linearly dependent, then every superset is linearly dependent.
- ⑰ If a set of vectors contain a 0 vector, then the set is always linearly dependent.
- ⑱ If you have more than n vectors in \mathbb{R}^n , then the homogeneous system $Vu=0$ is always linearly dependent.
 \hookrightarrow more variables than equations means infinitely many solutions.
- ⑲ Basis of a vector space is a linearly independent subset of the vector space that spans the vector space.
- ⑳ $\text{Rank}(A)$ = cardinality of a basis of the vector space.
 \hookrightarrow no. of non-zero rows in RREF.

WEEKS 5-8

- ① $\text{nullity}(A)$ = no. of independent variables set independent variables to 1 to find a basis of null space
- ② For an $m \times n$ matrix A , $\text{rank}(A) + \text{nullity}(A) = n$
- ③ Linear transformation properties: ① $f(u+v) = f(u) + f(v)$
 ② $f(c \cdot u) = c \cdot f(u)$
- ④ $f: V \rightarrow W$; $\ker(f) = \{v \in V \mid f(v) = 0\} \rightarrow$ null space
 $\text{Im}(f) = \{w \in W \mid \exists v \in V \text{ for which } f(v) = w\} \rightarrow$ column space
- ⑤ f is injective iff $\ker(f) = 0$
 f is surjective iff $\text{Im}(f) = W$
- ⑥ Basis of $\ker(f) \rightarrow$ use basis of null space and basis of domain.
 Basis of $\text{Im}(f) \rightarrow$ use column of dependent variable (in the matrix representation) and basis of co-domain.
- ⑦ $\text{Rank}(f) = \dim(\text{Im}(f))$
 $\text{Nullity}(f) = \dim(\ker(f))$
 $\text{rank}(f) + \text{nullity}(f) = \dim(V)$
- ⑧ $T: V \rightarrow W$ is a linear transformation
 β_1, β_2 are ordered bases of V A = matrix of T w.r.t. β_1 and γ_1
 γ_1, γ_2 " " " " W B = " " " " β_2 and γ_2
- $B = QAP$, where $P \rightarrow$ expresses β_2 in terms of β_1
 $Q \rightarrow$ " γ_2 " " " γ_1
- ⑨ Similar matrices
 ① $PB = AP$ ③ $\det(B) = \det(A)$
 ② $\text{rank}(A) = \text{rank}(B)$ ④ $\text{trace}(A) = \text{trace}(B)$
- ⑩ Affine subspace \rightarrow translation of a vector subspace
- ⑪ Norm
 $\|\cdot\|: V \rightarrow \mathbb{R}$
 ① $\|u+v\| \leq \|u\| + \|v\| \quad \forall u, v \in V$
 ② $\|cu\| = |c| \|u\| \quad \forall c \in \mathbb{R}, \forall u \in V$
 ③ $\|u\| \geq 0 \quad \forall u \in V$; $\|u\| = 0$ iff $u = 0$
- example: length of vector $\text{len}(u, y) = \sqrt{(u, y)(u, y)} = \sqrt{u^2 + y^2}$
 $\text{len}(u, y, z) = \sqrt{u^2 + y^2 + z^2}$
- ⑫ Angle b/w vectors v and u
 $\cos(\theta) = \frac{u \cdot v}{\|u\| \times \|v\|}$, where $\|u\| = \sqrt{u \cdot u}$

⑫ Inner product $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$

① $\langle v, v \rangle > 0 \quad \forall v \in V \setminus \{0\}$

② $\langle v_1 + v_2, v_3 \rangle = \langle v_1, v_3 \rangle + \langle v_2, v_3 \rangle$

③ $\langle v_1, v_2 \rangle = \langle v_2, v_1 \rangle$

④ $\langle cv_1, v_2 \rangle = \langle v_1, cv_2 \rangle = c \langle v_1, v_2 \rangle$

⑬ v_1, v_2 are orthogonal iff $\langle v_1, v_2 \rangle = 0$

⑭ Orthonormal basis S , $S = \{v_1, v_2, \dots, v_n\} \subseteq V$ is orthonormal iff: $\langle v_i, v_j \rangle = 0 \quad \forall i, j \in \{1, 2, \dots, n\}, i \neq j$
 $\|v_i\| = 1 \quad \forall i \in \{1, 2, \dots, n\}$

⑮ $S = \{v_1, v_2, \dots, v_n\}$ is an orthonormal basis of V .
 $v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n, \quad v \in V$
 $c_i = \langle v, v_i \rangle$

⑯ Projection of v onto w , where $v \in V, w \in W, W \subseteq V$
 Let $\{v_1, v_2, \dots, v_n\}$ be an orthonormal basis for W
 $\text{proj}_W(v) = \sum_{i=1}^n \langle v, v_i \rangle v_i$

① $P_W(v) = v \quad \forall v \in W$

② $\text{Range}(P_W) = W$

③ $W^\perp = \{v \mid v \in V, \text{ such that } \langle v, w \rangle = 0 \quad \forall w \in W\}$ is the null space of P_W

④ $P_W^2 = P_W$

⑤ $\|P_W(v)\| \leq \|v\|$

⑰ Orthogonal transformation $T: V \rightarrow V$ iff: $\langle Tv, Tw \rangle = \langle v, w \rangle \quad \forall v, w \in V$
 \hookrightarrow preserves lengths and angles
 \hookrightarrow matrix A iff: $A^T A = A A^T = I$

WEEKS 9-11

① Partial Derivative

$$\frac{\partial f}{\partial x_i}(a) = \lim_{h \rightarrow 0} \frac{f(a + h e_i) - f(a)}{h}, \text{ where } \{e_1, e_2, \dots\} \text{ is standard ordered basis of } \mathbb{R}^n$$

② Directional Derivative

\hookrightarrow rate of change in the direction of the unit vector u .

$$f_u(a) = \lim_{h \rightarrow 0} \frac{f(a + hu) - f(a)}{h}$$

Properties:

$$\rightarrow (cf + g)_u(a) = cf_u(a) + g_u(a)$$

$$\rightarrow (fg)_u(a) = f_u(a)g(a) + f(a)g_u(a)$$

$$\rightarrow (f/g)_u(a) = \frac{f_u(a)g(a) - f(a)g_u(a)}{(g(a))^2}$$

③ To check for continuity of a function at a point, check the limit of that function along different curves at that point.

④ $\nabla f(x) = \left(\frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right)$ in \mathbb{R}^n

⑤ Directional derivative at point a in direction of u :

$$\nabla f_u(a) = \nabla f(a) \cdot u$$

⑥ $f_u \rightarrow$ directional derivative in the direction of u .

Steepest ascent \rightarrow when $u = \frac{\nabla f}{\|\nabla f\|}$

Steepest descent \rightarrow when $u = -\frac{\nabla f}{\|\nabla f\|}$

No change \rightarrow when $\langle u, \nabla f \rangle = 0$, i.e. orthogonal

⑦ Equation of tangent line at point (a, b) in direction of $u = (u_1, u_2)$

$$x(t) = a + tu_1 \quad ; \quad y(t) = b + tu_2 \quad ; \quad z(t) = f(a, b) + t f_u(a, b) \quad , \text{ where } t f_u(a, b) = t (\nabla f(a, b) \cdot u)$$

⑧ Tangent Plane :

$$z = f(a, b) + \frac{\partial f}{\partial x}(a, b)(x-a) + \frac{\partial f}{\partial y}(a, b)(y-b)$$

⑨ Hessian matrix

Let $f(x_1, x_2, \dots, x_n)$ be defined on D in \mathbb{R}^n

The Hessian matrix of f is defined as:

$$Hf = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} & \dots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

⑩ Classification of critical points using Hf

f in $\mathbb{R}^2 \Rightarrow$

If $\det(Hf(a)) > 0$ $\begin{cases} f_{xx}(a) > 0 \rightarrow \text{then } a - \text{local minima} \\ f_{xx}(a) < 0 \rightarrow \text{then } a - \text{local maxima} \end{cases}$
If $\det(Hf(a)) < 0 \rightarrow$ then a - saddle point
If $\det(Hf(a)) = 0 \rightarrow$ then inconclusive

f in $\mathbb{R}^3 \Rightarrow$

If $\det(Hf(a)) > 0$ AND $(f_{xx}f_{yy} - f_{xy}^2)(a) > 0$ AND $f_{xx}(a) > 0 \Rightarrow$ local minimum
If $\det(Hf(a)) < 0$ AND $(f_{xx}f_{yy} - f_{xy}^2)(a) > 0$ AND $f_{xx}(a) < 0 \Rightarrow$ local minimum
If $\det(Hf(a)) \neq 0$ AND cases 1 or 2 do not occur \Rightarrow saddle point
If $\det(Hf(a)) = 0 \Rightarrow$ inconclusive

