

1 Preliminaries

1.1 A review of linear algebra

Vector Space Let \mathbb{R} be the scalar field of real numbers. We consider only real vector spaces. Let V_n be a set. V_n is a vector space (also called a linear space) if it is equipped with two operations:

$$\begin{aligned} \text{scalar product} \quad & \mathbb{R} \times V_n \rightarrow V_n, \\ \text{vector addition} \quad & V_n \times V_n \rightarrow V_n, \end{aligned}$$

and it is closed under these two operations. That is, V_n is a vector space if $\forall \alpha, \beta \in \mathbb{R} \ \& \ \forall \mathbf{a}, \mathbf{b} \in V_n$,

$$\alpha \mathbf{a} + \beta \mathbf{b} \in V_n.$$

The vector space V_n is **n -dimensional** if we can find a basis $\{\mathbf{e}_1, \dots, \mathbf{e}_n\} \subset V_n$ such that for any $\mathbf{a} \in V_n$, we have a unique decomposition

$$\mathbf{a} = \sum_{i=1}^n a_i \mathbf{e}_i,$$

where $a_i \in \mathbb{R}$ ($i = 1, \dots, n$) are the components (coordinates) of vector \mathbf{a} under the basis $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$.

Tensor Space Let V_n (V_m) be n -dimensional (m -dimensional) vector space. A mapping $\mathbf{A} : V_n \rightarrow V_m$ is a tensor if \mathbf{A} is linear. That is, $\forall \alpha, \beta \in \mathbb{R} \ \& \ \forall \mathbf{a}, \mathbf{b} \in V_n$,

$$\mathbf{A}(\alpha \mathbf{a} + \beta \mathbf{b}) = \alpha \mathbf{A}(\mathbf{a}) + \beta \mathbf{A}(\mathbf{b}). \quad (1)$$

Let $\text{Lin}(V_n, V_m)$ be the collection of all linear mappings (i.e., tensors) with domain V_n and range V_m . For any $\alpha \in \mathbb{R}$ and any $\mathbf{A}_1, \mathbf{A}_2 \in \text{Lin}(V_n, V_m)$, define two operations

$$\begin{aligned} \text{scalar product} \quad & (\alpha \mathbf{A}_1)(\mathbf{a}) = \alpha \mathbf{A}_1(\mathbf{a}) \quad \forall \mathbf{a} \in V_n, \\ \text{vector addition} \quad & (\mathbf{A}_1 + \mathbf{A}_2)(\mathbf{a}) = \mathbf{A}_1(\mathbf{a}) + \mathbf{A}_2(\mathbf{a}) \quad \forall \mathbf{a} \in V_n. \end{aligned}$$

◆ **CLAIM:** For any $\alpha, \beta \in \mathbb{R}$ and any $\mathbf{A}_1, \mathbf{A}_2 \in \text{Lin}(V_n, V_m)$, $\alpha \mathbf{A}_1 + \beta \mathbf{A}_2$ is a linear mapping (from V_n to V_m).

The above claim implies that the set $\text{Lin}(V_n, V_m)$ is also a vector space.

Inner Product We equip a n -dimensional vector space V_n with a mapping $V_n \times V_n \rightarrow \mathbb{R}$, called inner product such that for any $\alpha, \beta \in \mathbb{R}$ and any $\mathbf{a}, \mathbf{b}, \mathbf{c} \in V_n$, the inner product is

1. Positive-definite: $\mathbf{a} \cdot \mathbf{a} \geq 0$; $\mathbf{a} \cdot \mathbf{a} = 0 \iff \mathbf{a} = 0$,
2. Linear: $\mathbf{a} \cdot (\alpha \mathbf{b} + \beta \mathbf{c}) = \alpha \mathbf{a} \cdot \mathbf{b} + \beta \mathbf{a} \cdot \mathbf{c}$,
3. Symmetric: $\mathbf{a} \cdot \mathbf{b} = \mathbf{b} \cdot \mathbf{a}$.

Geometric interpretations:

- Length of a vector: $|\mathbf{a}| = \sqrt{\mathbf{a} \cdot \mathbf{a}}$,
- Angle between two vectors: $\cos(\theta) = \frac{\mathbf{a} \cdot \mathbf{b}}{|\mathbf{a}| |\mathbf{b}|}$.

Euclidean Space \mathbb{R}^n For a n -dimensional vector space V_n equipped with an inner product, we can find an orthonormal basis $\{\mathbf{e}_i : i = 1, \dots, n\}$ such that for all $i, j = 1, \dots, n$,

$$\mathbf{e}_i \cdot \mathbf{e}_j = \delta_{ij} = \begin{cases} 1 & \text{if } i = j, \\ 0 & \text{if } i \neq j, \end{cases}$$

where δ_{ij} is called *kroncker delta*. With respect to this basis, for any vector $\mathbf{a} \in V_n$, we find its components (a_1, \dots, a_n) (or coordinates if a is a point in space)

$$\mathbf{a} = \sum_{i=1}^n a_i \mathbf{e}_i, \quad a_i = \mathbf{a} \cdot \mathbf{e}_i \in \mathbb{R} \quad \forall i = 1, \dots, n.$$

We can further identify the space V_n with the familiar Euclidean space \mathbb{R}^n . However, one shall keep in mind, \mathbb{R}^n , as a vector space equipped with an inner product, is more than a collection of arrays of real numbers. One should not think of a vector in \mathbb{R}^n as an array of real numbers unless we specify a basis or a frame.

Tensor Product For vectors $\mathbf{a} \in \mathbb{R}^n$ and $\mathbf{b} \in \mathbb{R}^m$, the tensor product $\mathbf{b} \otimes \mathbf{a}$ is a linear mapping:

$$\begin{aligned} \mathbf{b} \otimes \mathbf{a} : V_n &\rightarrow V_m \\ (\mathbf{b} \otimes \mathbf{a})(\mathbf{c}) &= (\mathbf{a} \cdot \mathbf{c})\mathbf{b} \quad \forall \mathbf{c} \in \mathbb{R}^n. \end{aligned}$$

◆ CLAIM: For any $\mathbf{a} \in \mathbb{R}^n$ and $\mathbf{b} \in \mathbb{R}^m$, the mapping $\mathbf{b} \otimes \mathbf{a}$ (from V_n to V_m) defined above is linear.

◆ CLAIM: Let $\{\mathbf{e}_i : i = 1, \dots, n\}$ be an orthonormal basis of \mathbb{R}^n and $\{\hat{\mathbf{e}}_p : p = 1, \dots, m\}$ be an orthonormal basis of \mathbb{R}^m . Show that

$$\{\hat{\mathbf{e}}_p \otimes \mathbf{e}_i : i = 1, \dots, n, p = 1, \dots, m\} \subset \text{Lin}(\mathbb{R}^n, \mathbb{R}^m)$$

forms a basis of the linear space $\text{Lin}(\mathbb{R}^n, \mathbb{R}^m)$.

Subspace of \mathbb{R}^n , Orthogonal Subspace A subset $M \subset \mathbb{R}^n$ is a subspace if $\forall \alpha, \beta \in \mathbb{R} \ \& \ \forall \mathbf{a}, \mathbf{b} \in M$,

$$\alpha \mathbf{a} + \beta \mathbf{b} \in M.$$

Let $M^\perp = \{\mathbf{b} : \mathbf{b} \cdot \mathbf{a} = 0 \ \forall \mathbf{a} \in M\}$.

◆ CLAIM: Show that M^\perp is a subspace of \mathbb{R}^n if M is a subspace.

Projection Theorem Let M be a subspace of \mathbb{R}^n . For any $\mathbf{x} \in \mathbb{R}^n$, we have

$$\mathbf{x} = \mathbf{y} + \mathbf{z} \text{ where } \mathbf{y} \in M, \mathbf{z} \in M^\perp.$$

The vector \mathbf{y} , \mathbf{z} are uniquely determined by \mathbf{x} .

◆ PROOF:

Transpose of a Tensor Let $\mathbf{A} \in \text{Lin}(\mathbb{R}^n, \mathbb{R}^m)$, $\{\mathbf{e}_i : i = 1, \dots, n\}$ be an orthonormal basis of \mathbb{R}^n and $\{\hat{\mathbf{e}}_p : p = 1, \dots, m\}$ be an orthonormal basis of \mathbb{R}^m . Then \mathbf{A} admits the following decomposition

$$\mathbf{A} = \sum_{p,i} A_{pi} \hat{\mathbf{e}}_p \otimes \mathbf{e}_i \quad \text{where } A_{pi} = \hat{\mathbf{e}}_p \cdot \mathbf{A}(\mathbf{e}_i) \quad \forall i = 1, \dots, n, p = 1, \dots, m.$$

Define

$$\begin{aligned} \mathbf{A}^T &: \mathbb{R}^m \rightarrow \mathbb{R}^n, \\ \mathbf{A}^T &= \sum_{p,i} A_{pi} \mathbf{e}_i \otimes \hat{\mathbf{e}}_p \in \text{Lin}(\mathbb{R}^m, \mathbb{R}^n). \end{aligned}$$

◆ CLAIM: For any $\mathbf{a} \in \mathbb{R}^n$ and $\mathbf{b} \in \mathbb{R}^m$,

$$\mathbf{b} \cdot \mathbf{A}(\mathbf{a}) = \mathbf{a} \cdot \mathbf{A}^T(\mathbf{b}).$$

Symmetric and Skew-symmetric Tensor Let $\mathbf{A} \in \text{Lin}(\mathbb{R}^n, \mathbb{R}^n)$. \mathbf{A} is symmetric if $\mathbf{A} = \mathbf{A}^T$; \mathbf{A} is skew-symmetric if $\mathbf{A}^T = -\mathbf{A}$.

Let $\{\mathbf{e}_i : i = 1, \dots, n\}$, $\{\hat{\mathbf{e}}_p : p = 1, \dots, n\}$ be two orthonormal bases of \mathbb{R}^n . We have shown

$$\mathbf{A} = \sum_{p,i} A_{pi} \hat{\mathbf{e}}_p \otimes \mathbf{e}_i \quad \text{where } A_{pi} = \hat{\mathbf{e}}_p \cdot \mathbf{A}(\mathbf{e}_i) \quad \forall p, i = 1, \dots, n.$$

◆ CLAIMS :

1. For any $\mathbf{A} \in \text{Lin}(\mathbb{R}^n, \mathbb{R}^n)$, we have a unique decomposition $\mathbf{A} = \mathbf{E} + \mathbf{W}$, where $\mathbf{E} = \mathbf{E}^T$ and $\mathbf{W} = -\mathbf{W}^T$.
2. $\mathbf{A} = \mathbf{A}^T$ if and only if for any $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$,

$$\mathbf{b} \cdot \mathbf{A}(\mathbf{a}) = \mathbf{a} \cdot \mathbf{A}(\mathbf{b}).$$

3. If $\mathbf{A} = \mathbf{A}^T$ and $\mathbf{a} \cdot \mathbf{A}(\mathbf{a}) = 0$ for any $\mathbf{a} \in \mathbb{R}^n$, then $\mathbf{A} = 0$.
4. There exists a nonzero tensor \mathbf{A} such that

$$\mathbf{a} \cdot \mathbf{A}\mathbf{a} = 0 \quad \forall \mathbf{a} \in \mathbb{R}^n, n \geq 2.$$

5. Assume that $(\hat{\mathbf{e}}_1, \dots, \hat{\mathbf{e}}_n) = (\mathbf{e}_1, \dots, \mathbf{e}_n)$. If $\mathbf{A} = \mathbf{A}^T$, then $A_{pi} = A_{ip}$ for all $p, i = 1, \dots, n$; if $\mathbf{A} = -\mathbf{A}^T$, then $A_{pi} = -A_{ip}$.

Product of tensors Let $\mathbf{A} \in \text{Lin}(\mathbb{R}^n, \mathbb{R}^m)$, $\mathbf{B} \in \text{Lin}(\mathbb{R}^m, \mathbb{R}^k)$. Then

$$\begin{aligned} \mathbf{B}\mathbf{A} &: \mathbb{R}^n \rightarrow \mathbb{R}^k, \\ \mathbf{B}\mathbf{A}(\mathbf{a}) &= \mathbf{B}(\mathbf{A}(\mathbf{a})). \end{aligned}$$

Orthogonal Tensor Let $\mathbf{Q} \in \text{Lin}(\mathbb{R}^n, \mathbb{R}^n)$. The tensor \mathbf{Q} is orthogonal if $\mathbf{Q}\mathbf{a} \cdot \mathbf{Q}\mathbf{b} = \mathbf{a} \cdot \mathbf{b}$ for all $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$. From the definition we see that orthogonal tensor operating on vectors preserves the length of a vector and the angle between two vectors since

1. $|\mathbf{a}| = |\mathbf{Qa}|$, and

2. $\mathbf{a} \cdot \mathbf{b} = \mathbf{Qa} \cdot \mathbf{Qb}$.

◆ CLAIM: A tensor $\mathbf{Q} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is orthogonal if and only if $\mathbf{Q}^T \mathbf{Q} = \mathbf{Q} \mathbf{Q}^T = \mathbf{I}$, where \mathbf{I} is the identity mapping from \mathbb{R}^n to \mathbb{R}^n .

Trace and determinant of a tensor Let $\mathbf{A} \in \text{Lin}(\mathbb{R}^n, \mathbb{R}^n)$ and $\{\mathbf{e}_i : i = 1, \dots, n\}$ be an orthonormal basis. Then we have $\mathbf{A} = \sum_{p,i} A_{pi} \mathbf{e}_p \otimes \mathbf{e}_i$ and refer to $\text{Tr}(\mathbf{A}) = \sum_{p=1}^n A_{pp}$ as the trace of \mathbf{A} , $\det A = \det[A_{pi}]$ as the determinant of \mathbf{A} .

◆ CLAIM $\text{Tr}, \det : \text{Lin}(\mathbb{R}^n, \mathbb{R}^n) \rightarrow \mathbb{R}$ is independent of the choice of basis.

Rigid Rotation Tensor An orthogonal tensor $\mathbf{R} \in \text{Lin}(\mathbb{R}^n, \mathbb{R}^n)$ is a rigid rotation if $\det \mathbf{R} = +1$.
Representation theorem: For any $\mathbf{A} \in \text{Lin}(\mathbb{R}^n, \mathbb{R})$, there is an $\mathbf{a} \in \mathbb{R}^n$ such that Explicitly, if we have

$$\mathbf{A} = \sum_i A_{1i} \hat{\mathbf{e}}_1 \otimes \mathbf{e}_i, \quad \hat{\mathbf{e}}_1 = \mathbf{1},$$

then

$$\mathbf{a} = \sum_{i=1}^n A_{1i} \mathbf{e}_i.$$

Cross product in \mathbb{R}^3 For $\mathbf{a}, \mathbf{b} \in \mathbb{R}^3$,

$$\mathbf{a} \wedge \mathbf{b} = \begin{vmatrix} \mathbf{e}_1 & \mathbf{e}_2 & \mathbf{e}_3 \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{vmatrix} = \mathbf{W}(\mathbf{b}),$$

where $\mathbf{W} = \sum_{p,i} W_{pi} \mathbf{e}_p \otimes \mathbf{e}_i$,

$$[W_{p,i}] = \begin{pmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{pmatrix}.$$

◆ CLAIM: The following properties of cross products holds:

1. $\mathbf{b} \wedge \mathbf{a} = -\mathbf{a} \wedge \mathbf{b}$, $\mathbf{a} \cdot (\mathbf{a} \wedge \mathbf{b}) = 0$, $\mathbf{b} \cdot (\mathbf{a} \wedge \mathbf{b}) = 0$.

2. $(\mathbf{a} \wedge \mathbf{b}) \cdot \mathbf{c} = (\mathbf{b} \wedge \mathbf{c}) \cdot \mathbf{a} = (\mathbf{c} \wedge \mathbf{a}) \cdot \mathbf{b}$.

3. Geometric interpretation: show that $|\mathbf{a} \wedge \mathbf{b}|$ = area of the parallelogram formed by \mathbf{a} and \mathbf{b} ;
 $|\mathbf{c} \cdot (\mathbf{a} \wedge \mathbf{b})|$ = volume of the parallelepiped formed by $\mathbf{a}, \mathbf{b}, \mathbf{c}$.