

Recap: Last class we discussed
vector spaces,
dimension of a vector space
basis.

Then we introduced an additional structure called the inner product.

A useful identity about the inner product:

$$\boxed{\langle \underline{u}, \underline{v} \rangle \leq \|\underline{u}\| \|\underline{v}\|,} \quad \text{--- ①}$$

where $\|\underline{u}\| := \sqrt{\langle \underline{u}, \underline{u} \rangle}$ is called the norm of \underline{u} .

Equation ① is called the Cauchy-Schwarz inequality
(see note on Piazza for a proof).

Clarification: The word Euclidean space in some books ~~is some books~~ is used to refer to the Euclidean vector space, while in others it refers to the Euclidean point space. We will avoid this ambiguity by always referring explicitly to either "Euclidean VS" or "Euclidean point space".

We then discussed affine space, with the Euclidean point space as an example.

Tensor: A tensor is a linear transformation from one vector space to another. It is represented as

$$T: V \rightarrow W, \text{ where}$$

V, W are vector spaces.

what does "linear transformation" mean?

$$T(\alpha \underline{u} + \beta \underline{v}) = \alpha T\underline{u} + \beta T\underline{v} \quad \forall \alpha, \beta \in \mathbb{R}, \underline{u}, \underline{v} \in V$$

Because of linear transformation, property

$(T\underline{u} = \alpha_i T\underline{e}_{\sim i}, \text{ where } \underline{u} = \alpha_i \underline{e}_{\sim i})$, T is completely described by its action on the basis, i.e. $\{T\underline{e}_{\sim 1}, T\underline{e}_{\sim 2}, T\underline{e}_{\sim 3}\}$.

Typically, we deal with situations where $W = V$.

Therefore, T is completely described by

$$T_{ij} := \langle T\underline{e}_{\sim j}, \underline{e}_{\sim i} \rangle.$$

Therefore, T can be represented as a matrix.

But this representation depends on the choice of the basis. We will come back to the choice of

basis later. For now, we will always assume we have a fixed orthonormal basis

What does $T_{\underline{u}}$ look like in indicial notation? (3)

$T_{\underline{u}}$ is a vector, and for a given basis, let its components be $\alpha_1, \alpha_2, \alpha_3$:

$$T_{\underline{u}} = \alpha_1 \underline{e}_1 + \alpha_2 \underline{e}_2 + \alpha_3 \underline{e}_3.$$

Each component can be obtained as

$$\alpha_i = T_{\underline{u}} \cdot \underline{e}_i.$$

Since \underline{u} can be represented as $\underline{u} = u_j \underline{e}_j$,

$$\alpha_i = T(\underline{u} \cdot \underline{e}_j) \cdot \underline{e}_i$$

$$= u_j T_{\underline{e}_j} \cdot \underline{e}_i$$

$$= u_j T_{ij} \quad (\text{By the definition of } T_{ij})$$

Therefore, the i th component of $T_{\underline{u}}$ is given by $T_{ij} u_j$. In other words, $T_{\underline{u}}$ may be viewed as the product of a matrix and a vector!

We will now put together some facts resulting from ④ the linearity property of a tensor.

1) A second-order tensor maps the zero vector into itself

$$T \underline{0} = \underline{0}$$

Two second-order tensors S and T are equal if they

$$S \underline{u} = T \underline{u} \quad \forall \underline{u} \in V.$$

The set of all second-order tensors is denoted by Lin .

2) Let $S, T \in \text{Lin}$. Then their sum $S+T$ defined as

$$(S+T) \underline{u} = S \underline{u} + T \underline{u},$$

is a second-order tensor. Similarly the identities αT (where $\alpha \in \mathbb{R}$), $\underline{0}$ and \underline{I} defined as

$$(\alpha T) \underline{u} := \alpha (T \underline{u}),$$

$$\underline{0} \underline{u} := \underline{0},$$

$$\underline{I} \underline{u} := \underline{u}$$

are all second-order tensors.

③ For any $T \in \text{Lin}$, $\lambda \in \mathbb{R}$

$$(i) 0T = 0,$$

$$(iv) 0 + T = T + 0 = T$$

$$(ii) \lambda 0 = 0$$

$$(v) T + (-T) = -T + T = 0,$$

$$(iii) 1T = T$$

$$\text{where } -T := (-1)T.$$

The subtraction of two second-order tensors is defined as $S - T := S + (-1)T$.

Theorem: The set of all second-order tensors is a real vector space.

Let us now look at a special kind of a second order tensor that is constructed from a pair of vectors. Given two vectors $\underline{u}, \underline{v} \in V$, the tensor $\underline{u} \otimes \underline{v}$ defined as

$$(\underline{u} \otimes \underline{v}) \underline{w} := (\underline{v} \cdot \underline{w}) \underline{u},$$

is called a dyadic product of \underline{u} and \underline{v} .

(convince yourself that it is indeed a tensor.)

(6)

Ex: Show that $\underline{u} \otimes \underline{v}$ is a tensor

Ex: The tensor product is homogeneous,

$$(i) (\alpha \underline{u}) \otimes (\beta \underline{v}) = (\alpha \beta) (\underline{u} \otimes \underline{v}),$$

and distributive with respect to addition

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

$$(\underline{u} + \underline{v}) \otimes \underline{w} = \underline{u} \otimes \underline{w} + \underline{v} \otimes \underline{w}$$

Ex: $\underline{u} \otimes \underline{v} \neq \underline{v} \otimes \underline{u}$... Not commutative

Recall that for a given basis, a tensor T is completely represented by $T_{ij} \cdot \underline{e}_i \otimes \underline{e}_j$. We also showed that the space of tensors itself formed a vector space. Can we construct a basis for Lin ? Before we do that, we have the following theorem

Theorem: For Any second-order tensor T that maps a Euclidean vector space to itself, has a representation

as
$$T = T_{ij} \underline{e}_i \otimes \underline{e}_j,$$

where $T_{ij} = T \underline{e}_j \cdot \underline{e}_i$.

Proof. It is enough to show that

$$T_{\underline{u}} = (T_{ij} e_i \otimes e_j) \underline{u}$$

We know that $T_{\underline{u}}$ is given by

$$(T_{\underline{u}})_i = T_{ij} u_j \quad \text{--- (*)}$$

On the other hand

$$\begin{aligned} [T_{ij} e_i \otimes e_j] \underline{u} &= T_{ij} (e_j \cdot \underline{u}) e_i \quad \boxed{\text{By the definition of } u \otimes v} \\ &= T_{ij} u_j e_i \end{aligned}$$

\Rightarrow its i th component is $T_{ij} u_j$, which is identical to (*).