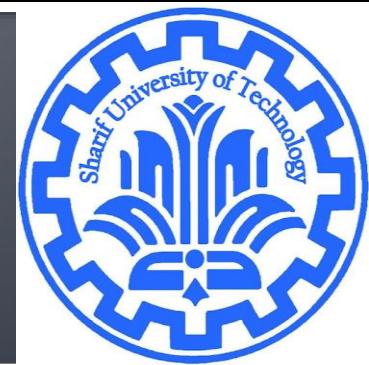


Introduction to Tensor

CE40282-1: Linear Algebra
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Sharif University of Technology



Covector

- Con~~vectors~~ (covariant vectors) are basically row vectors in an orthonormal basis, but not in general!

row sig

$\begin{bmatrix} 2 \\ 1 \end{bmatrix}$ \rightarrow $[2 \ 1]$

\vec{e}_2 \vec{e}_1

$\tilde{\vec{e}}_2$ $\tilde{\vec{e}}_1$

- Row vectors are function on column vectors.

$$\alpha: V \rightarrow \mathbb{R}$$
$$\alpha(\vec{v}) = \underbrace{\alpha_1 v^1 + \alpha_2 v^2 + \cdots + \alpha_n v^n}_{\sum \alpha_i v_i}$$

Covector

- A function that takes a vector and produces a scalar.
- Properties:

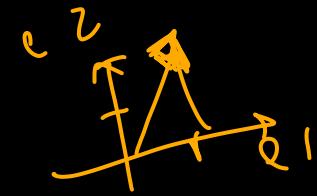
$$\alpha(\vec{v} + \vec{w}) = \alpha(\vec{v}) + \alpha(\vec{w})$$

$$\alpha(n\vec{v}) = n\alpha(\vec{v})$$

$$\alpha(n\vec{v} + m\vec{w}) = n\alpha(\vec{v}) + m\alpha(\vec{w})$$



Covector Visualization



$$\alpha(v) = [2 \ 1] \begin{pmatrix} x \\ y \end{pmatrix} = 2x + 1y$$

$2+2=4$

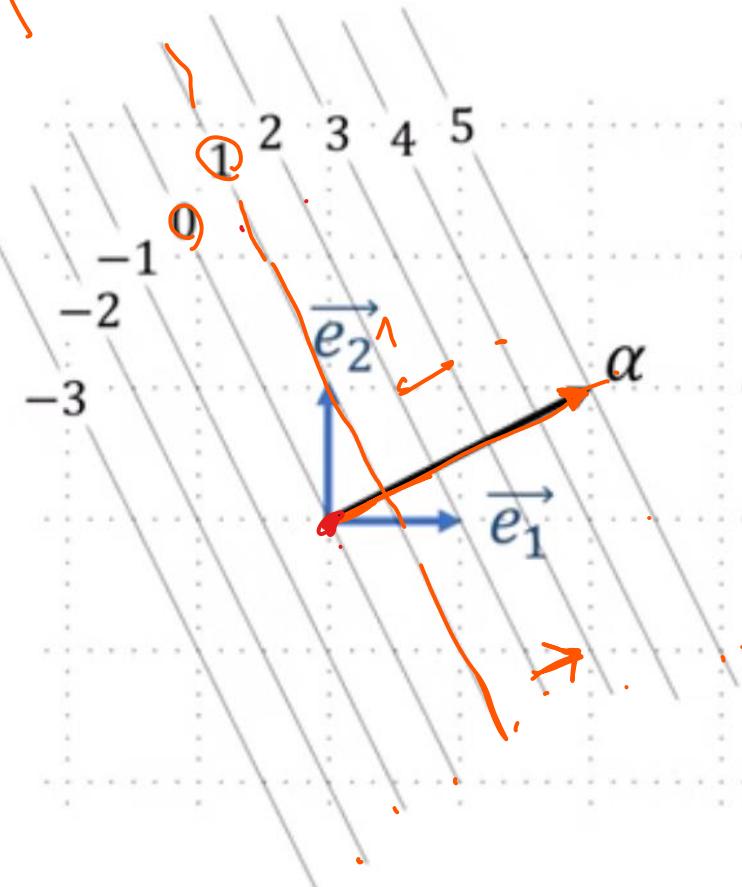
$$2x + 1y = 0 \rightarrow y = -2x$$

$$2x + 1y = 1 \rightarrow y = 1 - 2x$$

$$2x + 1y = 2$$

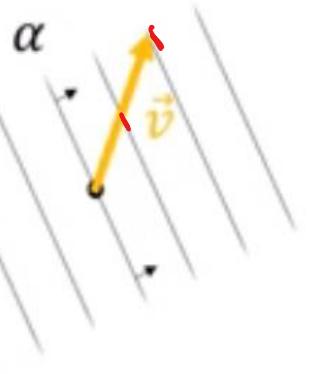
$$2x + 1y = -1$$

$$2x + 1y = -2$$

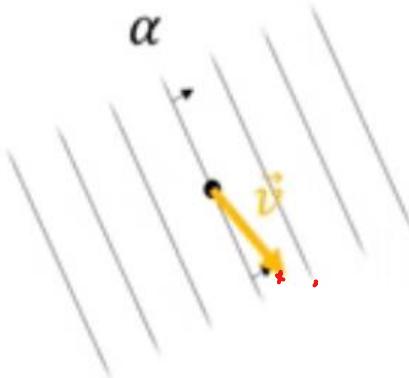


Covector Visualization

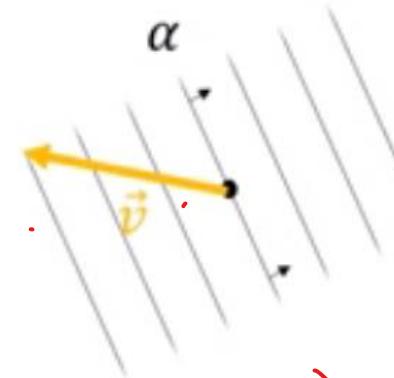
- Example of vector and covector



$$\alpha(r) = 2$$



$$\alpha(r) = 0.5$$



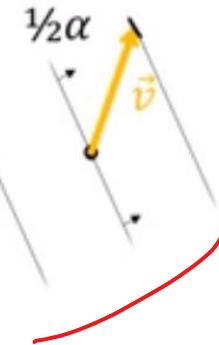
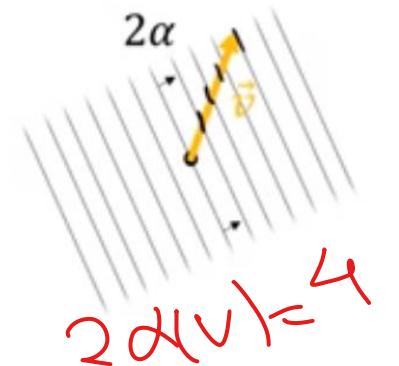
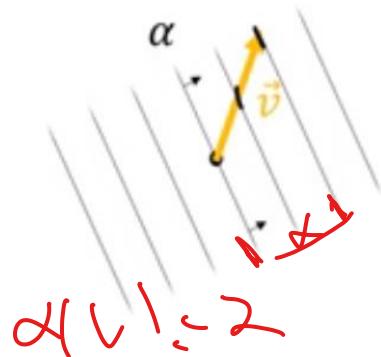
$$d(v) \Rightarrow 3$$

$$d(v) = \phi$$

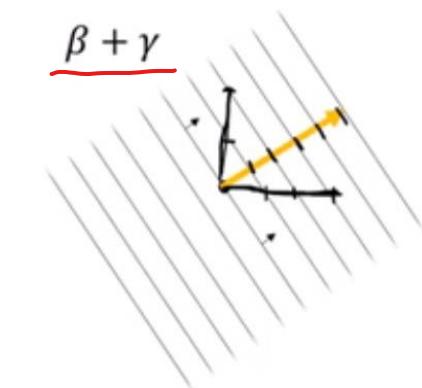
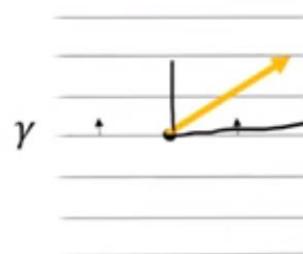
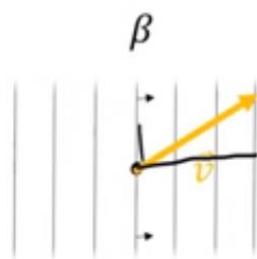
Covector Operations

$\alpha(\vec{v})$

■ Scaling



■ Adding



$$\underline{\beta(\vec{v}) = 3}$$

$$\underline{\gamma(\vec{v}) = 2}$$

$$(\beta + \gamma)(\vec{v}) = 5$$

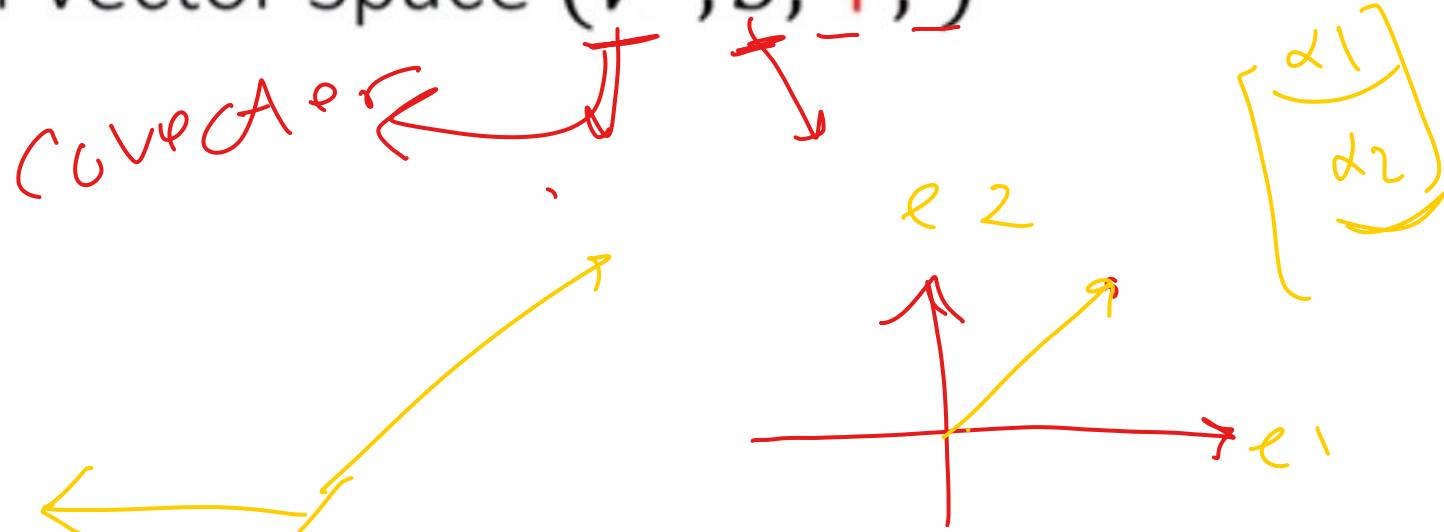
$$(\beta + \gamma)(\vec{v}) = \beta(\vec{v}) + \gamma(\vec{v})$$

Dual vector space

- Vector Space

$$(V, S, +, \cdot)$$

Dual Vector Space $(V^*, S, +, \cdot)$



Covector Components

When we write $\begin{bmatrix} 2 \\ 1 \end{bmatrix}_{\vec{e}_i}$, we mean $2\vec{e}_1 + 1\vec{e}_2$

What exactly are we measure with when we write $[2 \quad 1]$?

- Covectors are functions $\alpha: V \rightarrow \mathbb{R}$
- Covectors don't live in the vector space V
- We use can't basis vectors in V like $\{\vec{e}_1, \vec{e}_2\}$ to measure covectors

Take the basis $\{\vec{e}_1, \vec{e}_2\}$ for V .

Introduce two special covectors $\epsilon^1, \epsilon^2: V \rightarrow \mathbb{R}$

- $\epsilon^1(\vec{e}_1) = 1$ $\epsilon^1(\vec{e}_2) = 0$
- $\epsilon^2(\vec{e}_1) = 0$ $\epsilon^2(\vec{e}_2) = 1$

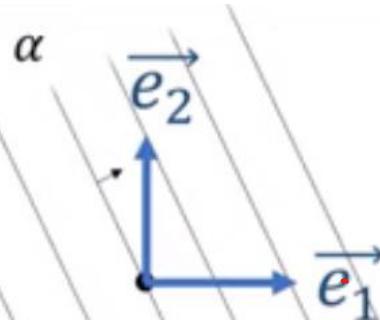
kron eck delta

$$\epsilon^i(\vec{e}_j) = \delta_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

Dual Basis

$$\alpha = \alpha_1 \epsilon^1 + \alpha_2 \epsilon^2$$

- $\alpha(\vec{e}_1) = \alpha_1$
- $\alpha(\vec{e}_2) = \alpha_2$

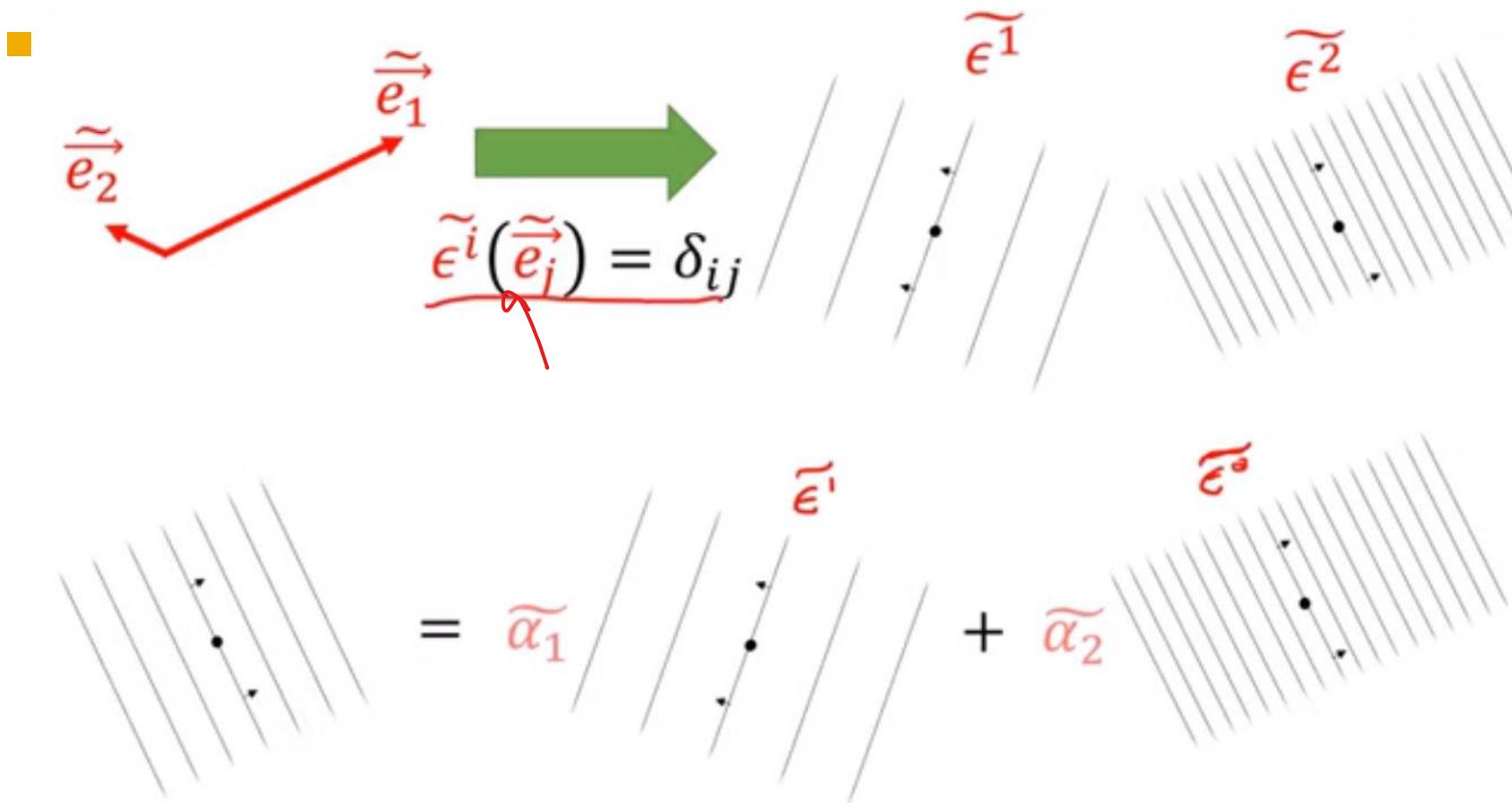


$\alpha(e_2)$

$$\alpha = \underbrace{\alpha_1}_{\text{d}(\epsilon^1)} \left| \begin{array}{c} \uparrow \\ \epsilon^1 \end{array} \right. + \underbrace{\alpha_2}_{\text{d}(\epsilon^2)} \left| \begin{array}{c} \epsilon^2 \\ \hline \dots \end{array} \right.$$

A diagram illustrating the decomposition of a vector α into a linear combination of dual basis elements ϵ^1 and ϵ^2 . The vector α is shown in a coordinate system with axes \vec{e}_1 and \vec{e}_2 . The decomposition is written as $\alpha = \alpha_1 \epsilon^1 + \alpha_2 \epsilon^2$. Red annotations highlight the coefficients α_1 and α_2 , and circled labels ϵ^1 and ϵ^2 identify the dual basis elements. A red arrow points from the bottom left towards the right side of the equation.

Dual Basis



Coordination change

$$\alpha = 2\epsilon^1 + 1\epsilon^2$$

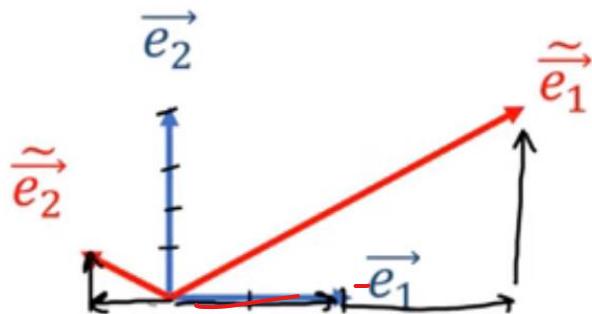
$$[2 \quad 1]_{\epsilon^i}$$

- $\alpha(\vec{e}_1) = \alpha_1 = 2$
- $\alpha(\vec{e}_2) = \alpha_2 = 1$

$$\alpha = \widetilde{\alpha}_1 \widetilde{\epsilon}^1 + \widetilde{\alpha}_2 \widetilde{\epsilon}^2$$

$$[\widetilde{\alpha}_1 \quad \widetilde{\alpha}_2]_{\widetilde{\epsilon}^i}$$

- $\alpha(\widetilde{\vec{e}}_1) = \widetilde{\alpha}_1$
- $\alpha(\widetilde{\vec{e}}_2) = \widetilde{\alpha}_2$



$$\begin{aligned}
 \widetilde{\alpha}_1 &= \alpha(\widetilde{\vec{e}}_1) \\
 &= \alpha(2\vec{e}_1 + 1\vec{e}_2) \\
 &= 2\alpha(\vec{e}_1) + 1\alpha(\vec{e}_2) \\
 &= 2(2) + 1(1) \\
 &= 5
 \end{aligned}$$

$$\begin{aligned}
 \widetilde{\alpha}_2 &= \alpha(\widetilde{\vec{e}}_2) \\
 &= \alpha(-\frac{1}{2}\vec{e}_1 + \frac{1}{4}\vec{e}_2) \\
 &= -\frac{1}{2}\alpha(\vec{e}_1) + \frac{1}{4}\alpha(\vec{e}_2) \\
 &= -\frac{1}{2}(2) + \frac{1}{4}(1) \\
 &= -\frac{3}{4}
 \end{aligned}$$

$$\left[\begin{array}{cc} \frac{1}{4} & \frac{1}{2} \\ -1 & 2 \end{array} \right] \left[\begin{array}{c} 2 \\ 1 \end{array} \right]_{\epsilon^i} = \left[\begin{array}{c} 1 \\ 0 \end{array} \right]_{\widetilde{\epsilon}^i}$$

$$[2 \quad 1]_{\epsilon^i}$$

$$B \left[\begin{array}{c} 2 \\ 1 \end{array} \right]_{\epsilon^i} = \left[\begin{array}{c} 1 \\ 0 \end{array} \right]_{\widetilde{\epsilon}^i}$$

$$[5 \quad -\frac{3}{4}]_{\widetilde{\epsilon}^i}$$

$$F \left[\begin{array}{c} 1 \\ 0 \end{array} \right]_{\widetilde{\epsilon}^i} = \left[\begin{array}{c} 2 \\ 1 \end{array} \right]_{\epsilon^i}$$

Conclusion

- Why “convector is a row vector” is not true?

$$\checkmark \begin{bmatrix} 2 \\ 1 \end{bmatrix}_{\epsilon^i} \leftrightarrow \begin{bmatrix} 2 & 1 \end{bmatrix}_{\epsilon^i}$$

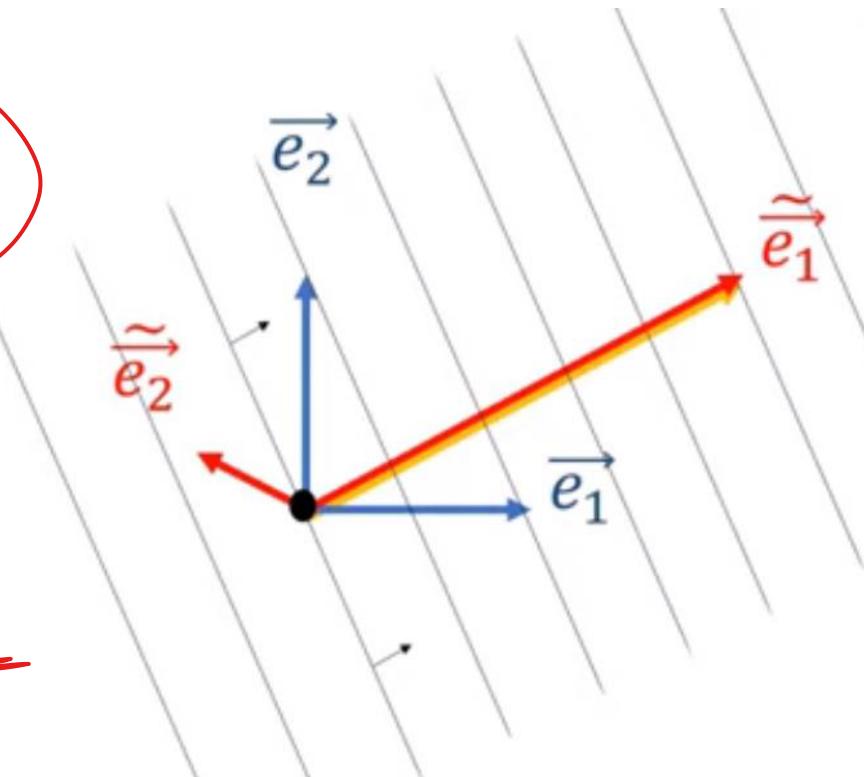
$\alpha(v)$

$$\checkmark \begin{bmatrix} 1 \\ 0 \end{bmatrix}_{\tilde{\epsilon}^i}$$

$\alpha(v)$

$$\begin{bmatrix} 5 & -\frac{3}{4} \end{bmatrix}_{\tilde{\epsilon}^i}$$

~~$\alpha(v)$~~



Kronecker Product

Definition 14.1 Let A be an $m \times n$ matrix and B a $p \times q$ matrix. The **Kronecker product** of A and B , written as $A \otimes B$, is the $mp \times nq$ partitioned matrix whose (i, j) -th block is the $p \times q$ matrix $a_{ij}B$ for $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$. Thus,

$$A \otimes B = \begin{bmatrix} a_{11}B & a_{12}B & \dots & a_{1n}B \\ a_{21}B & a_{22}B & \dots & a_{2n}B \\ \vdots & \vdots & & \vdots \\ a_{m1}B & a_{m2}B & \dots & a_{mn}B \end{bmatrix}.$$

■ Example

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 0 & 1 \end{bmatrix} \otimes \begin{bmatrix} 5 & 6 & 0 \\ 7 & 8 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 5 & 6 & 0 & 10 & 12 & 0 \\ 7 & 8 & 0 & 14 & 16 & 0 \\ 0 & 0 & 1 & 0 & 0 & 2 \\ 15 & 18 & 0 & 20 & 24 & 0 \\ 21 & 24 & 0 & 28 & 32 & 0 \\ 0 & 0 & 3 & 0 & 0 & 4 \\ 0 & 0 & 0 & 5 & 6 & 0 \\ 0 & 0 & 0 & 7 & 8 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Kronecker Product

Let A , B and C be any three matrices.

- - (i) $\alpha \otimes A = A \otimes \alpha = \alpha A$, for any scalar α .
 - (ii) $(\alpha A) \otimes (\beta B) = \alpha \beta (A \otimes B)$, for any scalars α and β .
 - (iii) $(A \otimes B) \otimes C = A \otimes (B \otimes C)$.
 - (iv) $(A + B) \otimes C = (A \otimes C) + (B \otimes C)$, if A and B are of same size.
 - (v) $A \otimes (B + C) = (A \otimes B) + (A \otimes C)$, if B and C are of same size.
 - (vi) $(A \otimes B)' = A' \otimes B'$.
- Other Properties:

$$A \otimes B \neq B \otimes A$$

$$\begin{matrix} \circ & \otimes & A = & A & \otimes & \circ \\ \circ & \otimes & A = & A & \otimes & \circ \end{matrix} = \textcircled{0}$$

$D \otimes A =$
skip

- Ref: Chapter 14: Linear Algebra and Matrix Analysis for Statistics

Kronecker Product

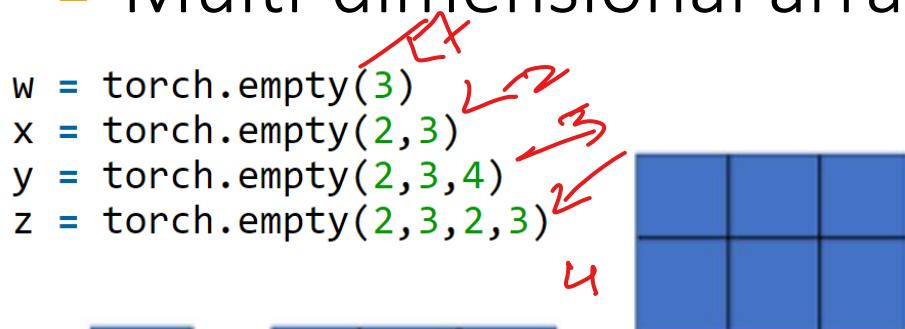
$$\begin{bmatrix} v^1 \\ v^2 \end{bmatrix} \otimes [\alpha_1 \quad \alpha_2]$$
$$\begin{bmatrix} [v^1] & \alpha_1 \\ [v^2] & \alpha_2 \end{bmatrix}$$
$$\begin{bmatrix} [v^1 \alpha_1] & [v^1 \alpha_2] \\ [v^2 \alpha_1] & [v^2 \alpha_2] \end{bmatrix}$$

$$\begin{bmatrix} [v^1 \alpha_1] & [v^1 \alpha_2] \\ [v^2 \alpha_1] & [v^2 \alpha_2] \end{bmatrix} \otimes [\omega_1 \quad \omega_2]$$
$$\begin{bmatrix} [[v^1 \alpha_1] & [v^1 \alpha_2]] & \omega_1 \\ [[v^2 \alpha_1] & [v^2 \alpha_2]] & \omega_2 \end{bmatrix} \rightarrow \begin{bmatrix} [[v^1 \alpha_1 \omega_1] & [v^1 \alpha_2 \omega_1]] \\ [[v^2 \alpha_1 \omega_1] & [v^2 \alpha_2 \omega_1]] \\ [[v^1 \alpha_1 \omega_2] & [v^1 \alpha_2 \omega_2]] \\ [[v^2 \alpha_1 \omega_2] & [v^2 \alpha_2 \omega_2]] \end{bmatrix}$$

Tensor

- Multi-dimensional array of numbers

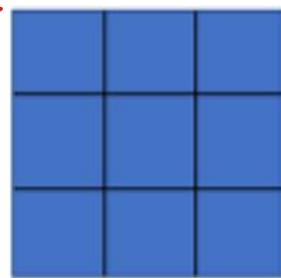
```
w = torch.empty(3)      ↗  
x = torch.empty(2, 3)    ↗  
y = torch.empty(2, 3, 4) ↗  
z = torch.empty(2, 3, 2, 3) ↗  
                                ↘
```



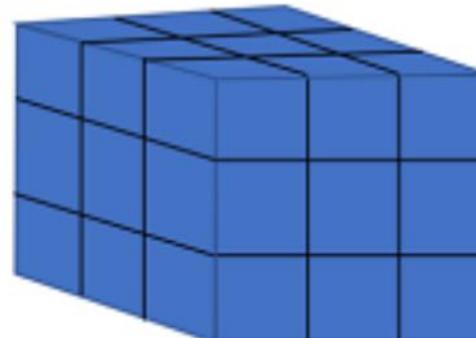
Scalar
(rank 0)



Vector
(rank 1)



Matrix
(rank 2)

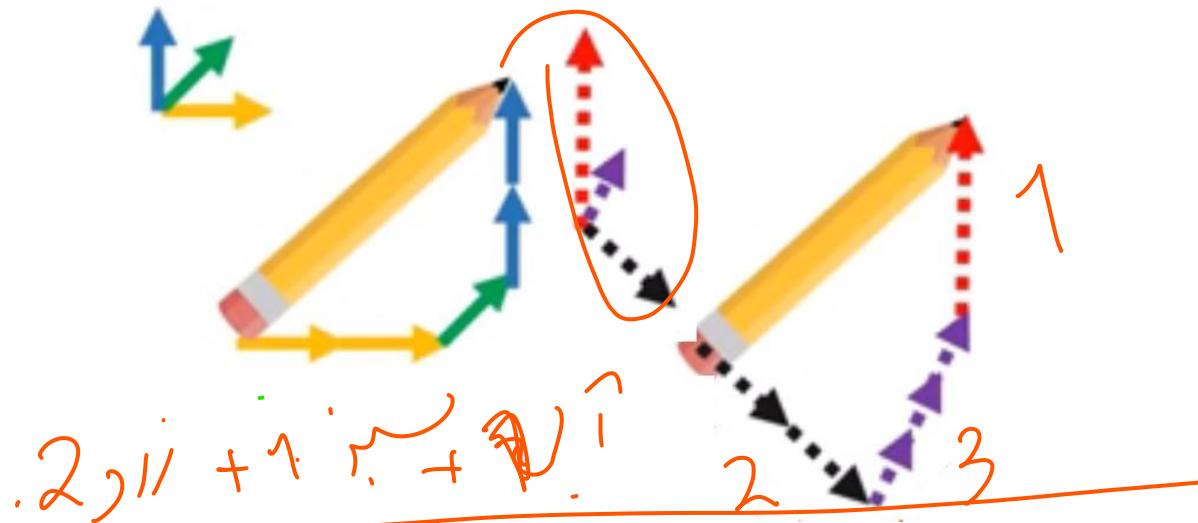
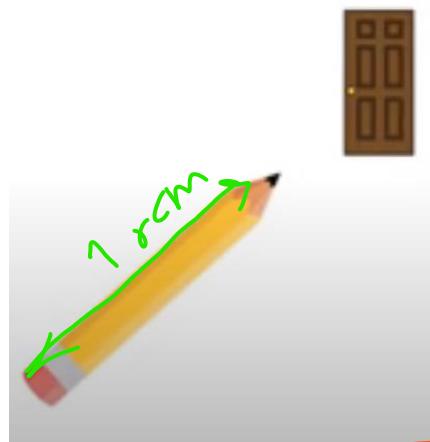


Rank-3 Tensor
(rank 3)

Tensor = an object that is invariant under a change of coordinates, and...
has components that change in a special, predictable way under a change of coordinates

Tensor

- Example!



Tensor = a collection of vectors and covectors combined together using the tensor product

Extra: Tensors as partial derivatives and gradients that transform with the Jacobian Matrix

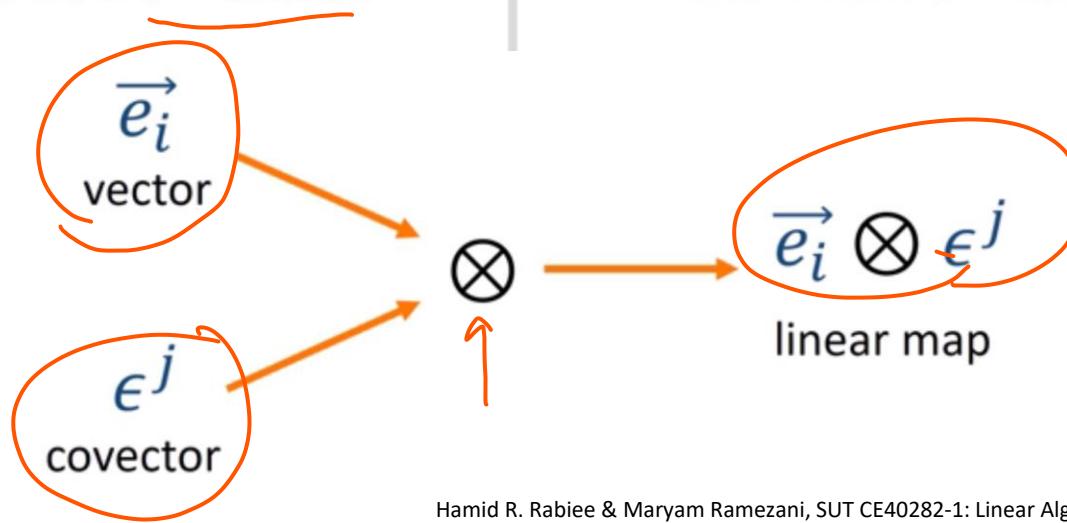
Tensor Product

The \otimes symbol...

Tensor Product
(Tensors)

$$\vec{e}_i \otimes \epsilon^j$$

Combines 2 tensors
into a new 3rd tensor.



Kronecker Product
(Arrays)

$$\begin{bmatrix} v^1 \\ v^2 \end{bmatrix} \otimes [\alpha_1 \quad \alpha_2]$$

Combines 2 arrays
into a new 3rd array.

Tensor Product

$$v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$$

$$\begin{aligned} & (\vec{e}_i \otimes e^j)(\vec{v}) \\ &= e_i \otimes (e^j(v)) = e_i \otimes (e^j(v_k e_k)) \\ &= v^k e_i \otimes (e^j(e_k)) \\ &= v^k e_i \delta_{jk} \end{aligned}$$

~~$v^i e_i$~~

Tensor vs Kronecker Product

Tensor Product

(Tensors)

$$\vec{v} \otimes \underline{\alpha} = (\vec{v}^i e_i) \otimes (\underline{\alpha_j} \epsilon^j)$$
$$= v^i \underline{\alpha_j} (\vec{e}_i \otimes \underline{\epsilon^j})$$

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

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

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

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

2x2

Kronecker Product

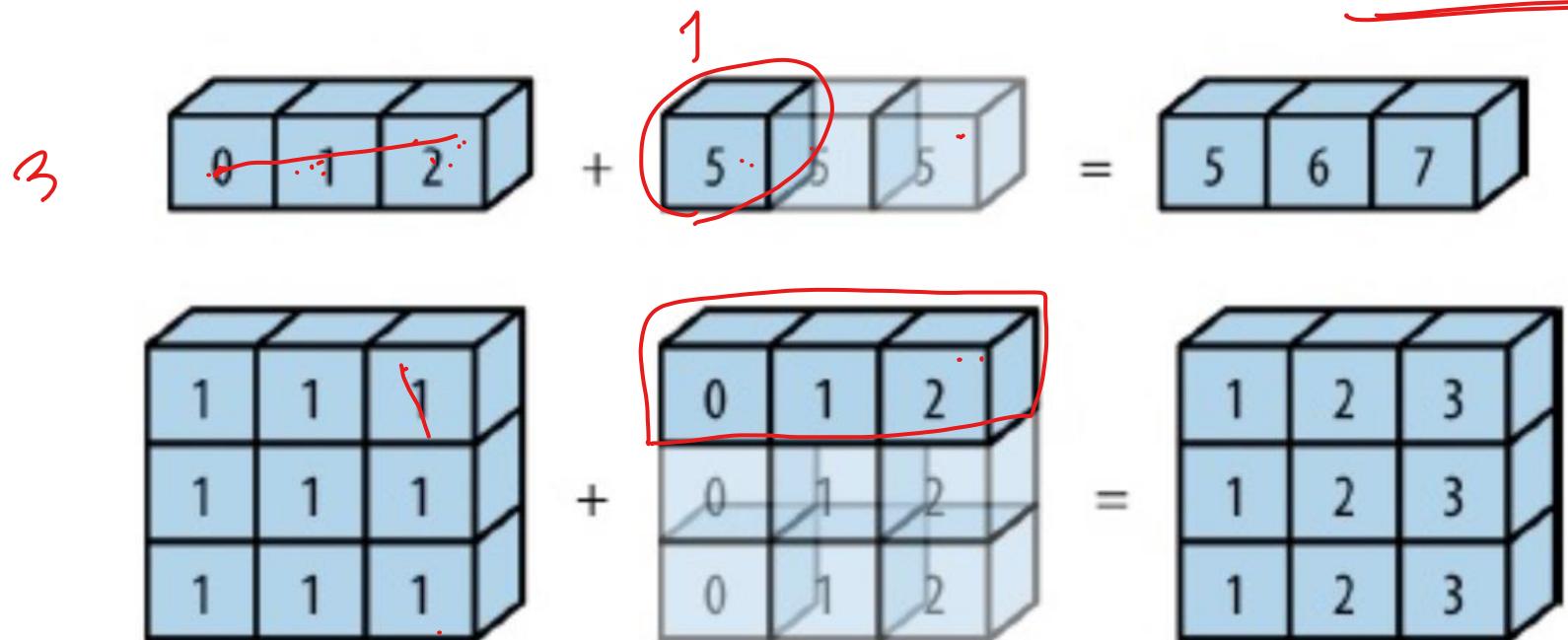
(Arrays)

$$\begin{bmatrix} v^1 \\ v^2 \end{bmatrix} \otimes [\alpha_1 \quad \alpha_2]$$

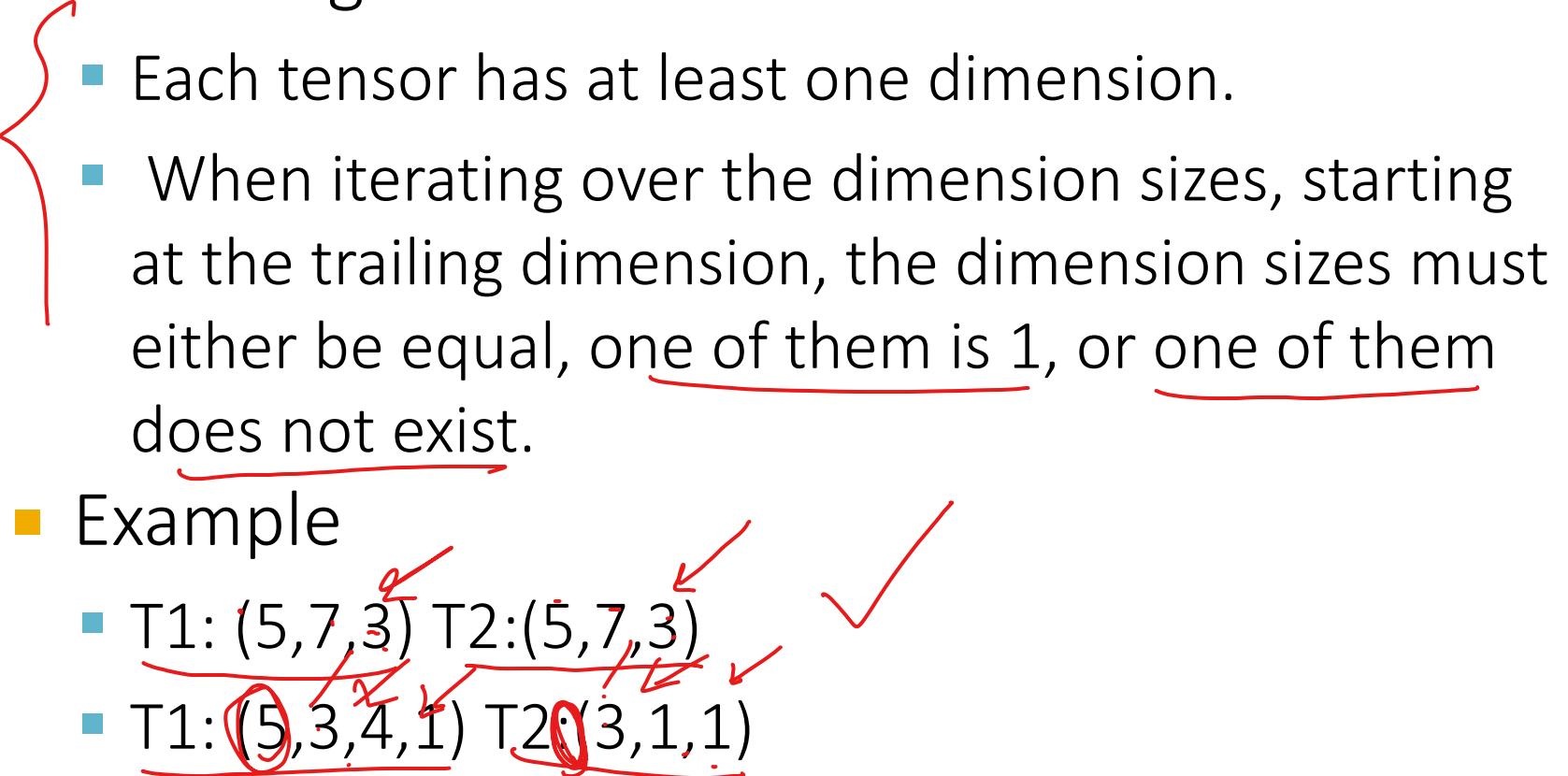
$$\begin{bmatrix} [v^1 \alpha_1] & [v^1 \alpha_2] \\ [v^2 \alpha_1] & [v^2 \alpha_2] \end{bmatrix}$$

Tensors Addition

- Adding tensors with same size ✓
- Adding scalar to tensor
- Adding tensors with different size: if broadcastable



Tensors Addition

- Two tensors are “broadcastable” if the following rules hold:
 - Each tensor has at least one dimension.
 - When iterating over the dimension sizes, starting at the trailing dimension, the dimension sizes must either be equal, one of them is 1, or one of them does not exist.
 - Example
 - T1: $(5, 7, 3)$ T2: $(5, 7, 3)$
 - T1: $(5, 3, 4, 1)$ T2: $(3, 1, 1)$
- 

Notes

■ Matrix Product on tensors

$$(m \times n) \cdot (n \times k) = (m \times k)$$

product is defined

The diagram illustrates the compatibility of dimensions for matrix multiplication. It shows two tensors being multiplied: one of size $(m \times n)$ and another of size $(n \times k)$. The middle dimension, n , is highlighted in red with a horizontal line underneath. A blue curved arrow points from the n in the first tensor to the n in the second tensor, indicating they must be equal for the product to be defined. Two red curved arrows point from the n in the first tensor to the k in the second tensor, one from below and one from above, indicating they must be compatible for multiplication. The resulting dimension of the product, $(m \times k)$, is shown in blue.



Derivative of a vector with respect to a matrix

$$r = \begin{bmatrix} r_1 \\ r_2 \end{bmatrix}_{m \times 1} \quad W = \begin{bmatrix} w_1 & w_2 & w_3 \\ w_4 & w_5 & w_6 \end{bmatrix}_{n \times k}$$

$\frac{\partial r}{\partial W} =$

$$= \begin{bmatrix} \frac{\partial r_1}{\partial w_1} & \frac{\partial r_1}{\partial w_2} & \dots & \frac{\partial r_1}{\partial w_k} \\ \frac{\partial r_2}{\partial w_1} & \frac{\partial r_2}{\partial w_2} & \dots & \frac{\partial r_2}{\partial w_k} \end{bmatrix}_{m \times (nk)}$$

$\frac{\partial r_1}{\partial w_1}$

m

$n \times n \times k$

Derivative of a matrix with respect to a matrix

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$m \times n$

$$\frac{\partial A}{\partial w}$$

$$w \in k \times l$$

$$\frac{\partial a_{11}}{\partial w}$$

$$\frac{\partial a_{12}}{\partial w}$$

$$\frac{\partial a_{21}}{\partial w}$$

$$\frac{\partial a_{22}}{\partial w}$$

$$\frac{\partial a_{11}}{\partial v}$$

$$\frac{\partial a_{12}}{\partial v}$$

$$\frac{\partial a_{21}}{\partial v}$$

$$\frac{\partial a_{22}}{\partial v}$$

