

# TENSOR ALGEBRA

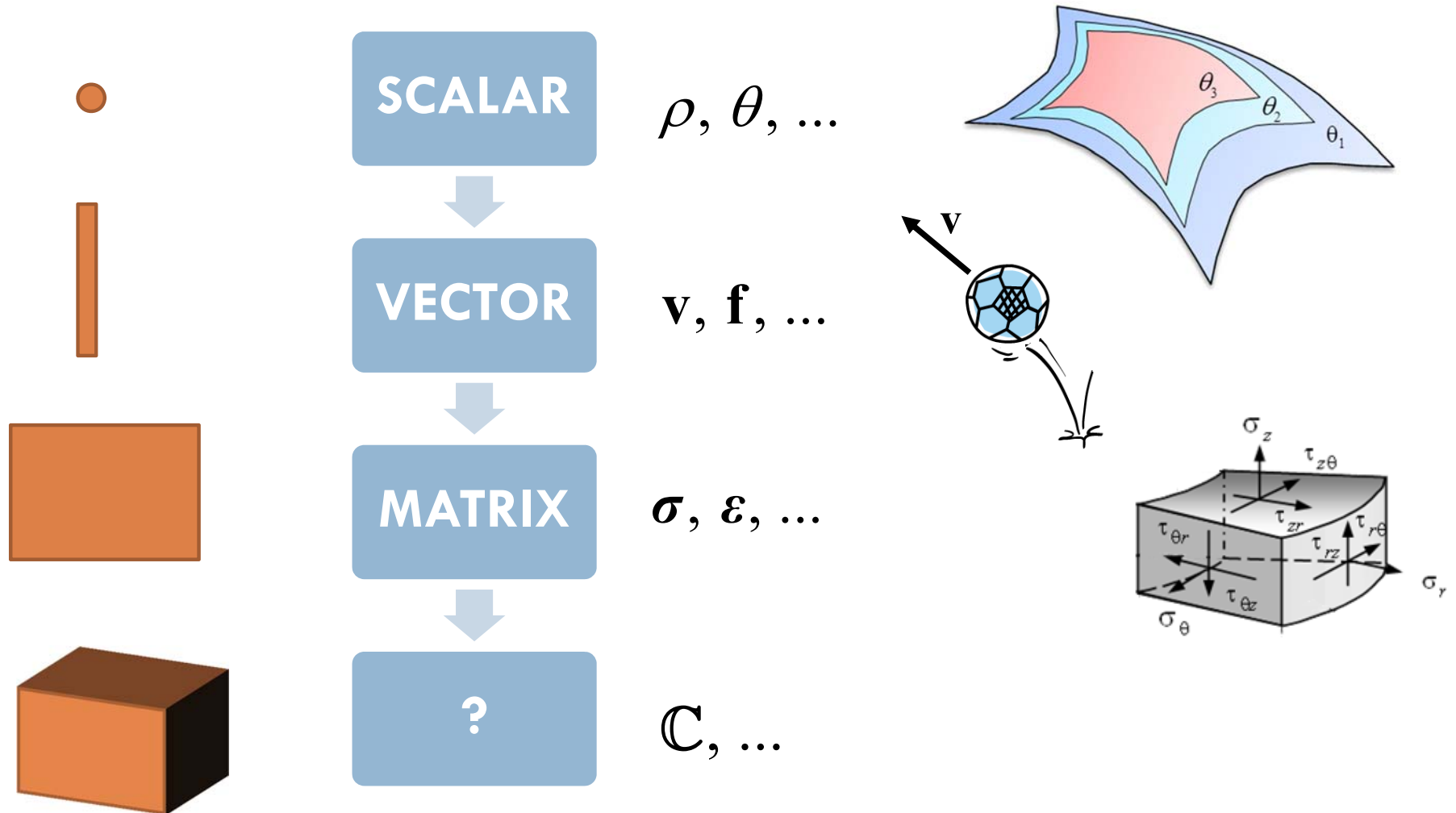
Continuum Mechanics Course (MMC) - ETSECCPB - UPC



# Introduction to Tensors

## Tensor Algebra

# Introduction



# Concept of Tensor

- ▣ A **TENSOR** is an algebraic entity with various components which **generalizes the concepts** of scalar, vector and matrix.
- Many physical quantities are mathematically represented as tensors.
- **Tensors are independent of any reference system** but, by need, are commonly represented in one by means of their “component matrices”.
- The **components** of a tensor **will depend on the reference system** chosen and will vary with it.

# Order of a Tensor

- The order of a tensor is given by the number of indexes needed to specify without ambiguity a component of a tensor.

$a$  ■ **Scalar**: zero dimension  $\alpha = 3.14$

$\underline{a}$ ,  $\mathbf{a}$  ■ **Vector**: 1 dimension

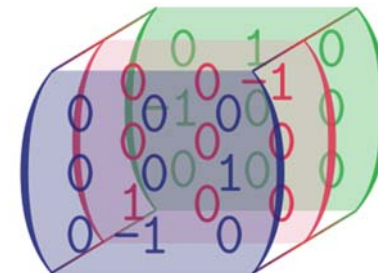
$\underline{\underline{A}}$ ,  $\mathbf{A}$  ■ **2<sup>nd</sup> order**: 2 dimensions

$\mathcal{A}$ ,  $\mathcal{A}$  ■ **3<sup>rd</sup> order**: 3 dimensions

$\mathbb{A}$ ,  $\mathbb{A}$  ■ **4<sup>th</sup> order** ...

$$v_i = \begin{pmatrix} 1.2 \\ 0.3 \\ 0.8 \end{pmatrix}$$

$$E_{ij} = \begin{pmatrix} 0.1 & 0 & 1.3 \\ 0 & 2.4 & 0.5 \\ 1.3 & 0.5 & 5.8 \end{pmatrix}$$

$$\epsilon_{ijk} =$$


# Cartesian Coordinate System

- Given an **orthonormal basis** formed by three mutually perpendicular unit vectors:

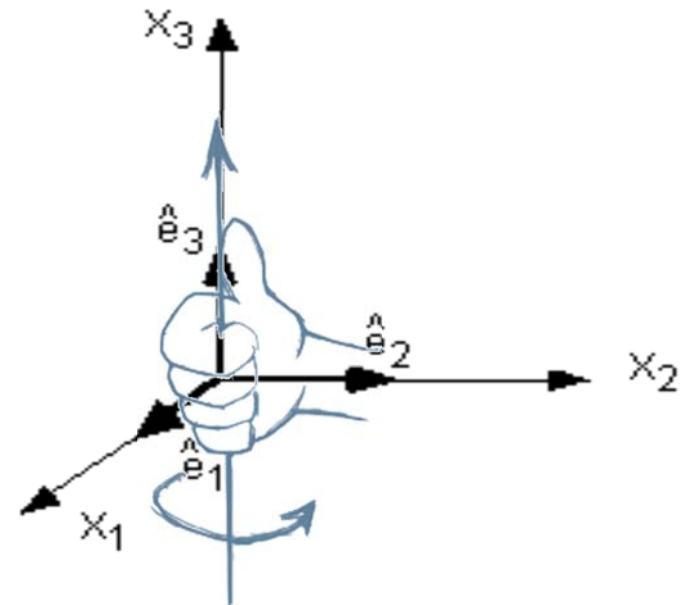
$$\hat{\mathbf{e}}_1 \perp \hat{\mathbf{e}}_2 , \quad \hat{\mathbf{e}}_2 \perp \hat{\mathbf{e}}_3 , \quad \hat{\mathbf{e}}_3 \perp \hat{\mathbf{e}}_1$$

Where:

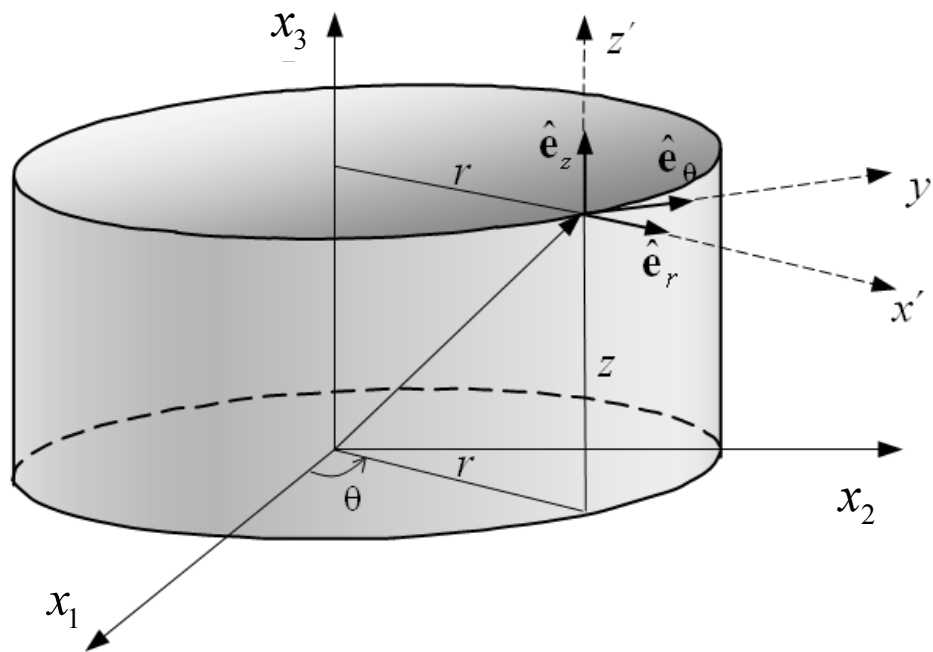
$$|\hat{\mathbf{e}}_1| = 1 , \quad |\hat{\mathbf{e}}_2| = 1 , \quad |\hat{\mathbf{e}}_3| = 1$$

■ Note that

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



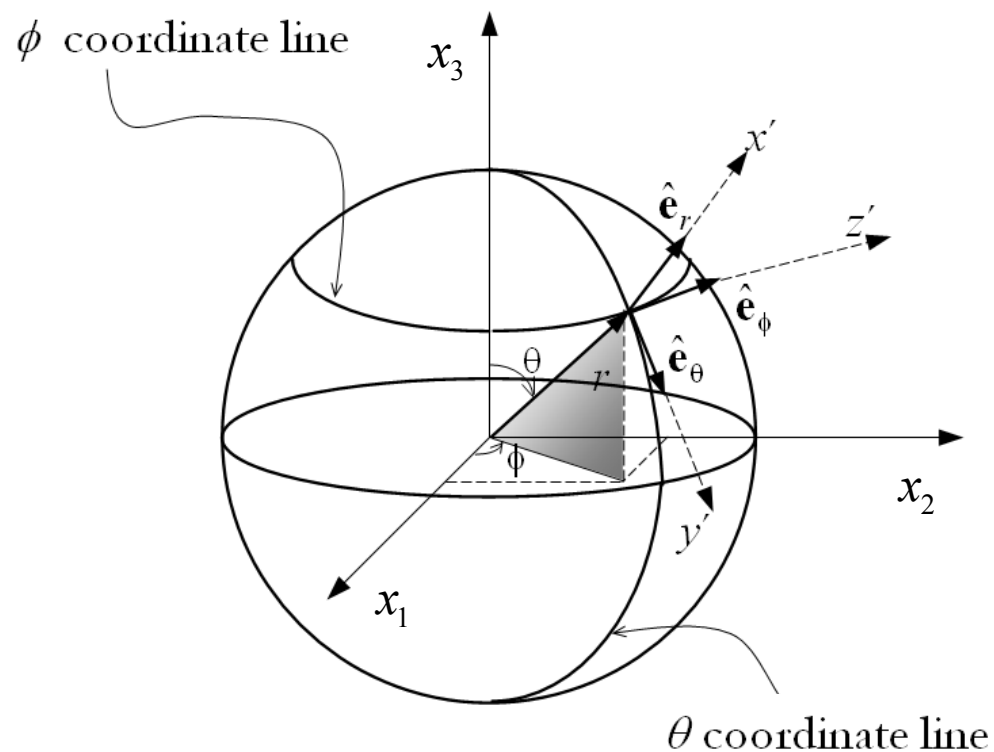
# Cylindrical Coordinate System



$$\mathbf{x}(r, \theta, z) \equiv \begin{cases} x_1 = r \cos \theta \\ x_2 = r \sin \theta \\ x_3 = z \end{cases}$$

$$\begin{aligned} \hat{e}_r &= \cos \theta \hat{e}_1 + \sin \theta \hat{e}_2 \\ \hat{e}_\theta &= -\sin \theta \hat{e}_1 + \cos \theta \hat{e}_2 \\ \hat{e}_z &= \hat{e}_3 \end{aligned}$$

# Spherical Coordinate System



$$\mathbf{x}(r, \theta, \phi) \equiv \begin{cases} x_1 = r \sin \theta \cos \phi \\ x_2 = r \sin \theta \sin \phi \\ x_3 = r \cos \theta \end{cases}$$

$$\begin{aligned} \hat{\mathbf{e}}_r &= \sin \theta \sin \phi \hat{\mathbf{e}}_1 + \sin \theta \cos \phi \hat{\mathbf{e}}_2 + \cos \theta \hat{\mathbf{e}}_3 \\ \hat{\mathbf{e}}_\theta &= \cos \phi \hat{\mathbf{e}}_1 - \sin \phi \hat{\mathbf{e}}_2 \\ \hat{\mathbf{e}}_\phi &= \cos \theta \sin \phi \hat{\mathbf{e}}_1 + \cos \theta \cos \phi \hat{\mathbf{e}}_2 - \sin \theta \hat{\mathbf{e}}_3 \end{aligned}$$





# Indicial or (Index) Notation

Tensor Algebra

# Tensor Bases – VECTOR

- A vector  $\mathbf{v}$  can be written as a unique linear combination of the **three** vector basis  $\hat{\mathbf{e}}_i$  for  $i \in \{1, 2, 3\}$ .

$$\mathbf{v} = v_1 \hat{\mathbf{e}}_1 + v_2 \hat{\mathbf{e}}_2 + v_3 \hat{\mathbf{e}}_3$$

- In matrix notation:

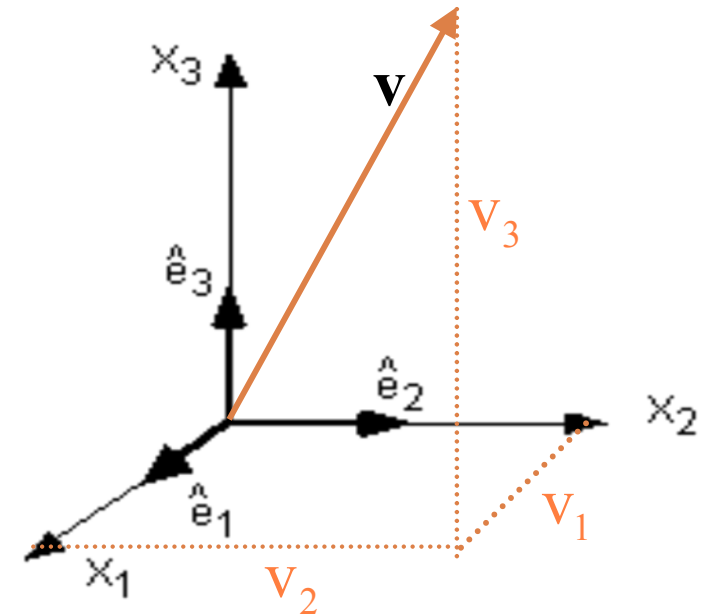
$$[\mathbf{v}] = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$$

- In index notation:

$$\mathbf{v} = \sum_i v_i \hat{\mathbf{e}}_i$$
$$[\mathbf{v}]_i = v_i$$

tensor as a physical entity

component  $i$  of the tensor in the given basis  $i \in \{1, 2, 3\}$



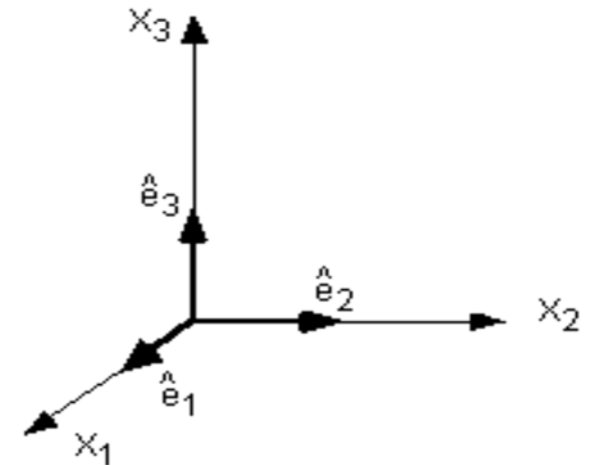
# Tensor Bases – 2<sup>nd</sup> ORDER TENSOR

- A 2<sup>nd</sup> order tensor  $\mathbf{A}$  can be written as a unique linear combination of the **nine** dyads  $\hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \equiv \hat{\mathbf{e}}_i \hat{\mathbf{e}}_j$  for  $i, j \in \{1, 2, 3\}$ .

$$\begin{aligned}\mathbf{A} = & A_{11} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + A_{12} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2) + A_{13} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_3) + \\ & + A_{21} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + A_{22} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2) + A_{23} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_3) + \\ & + A_{31} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + A_{32} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_2) + A_{33} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_3)\end{aligned}$$

Alternatively, this could have been written as:

$$\begin{aligned}\mathbf{A} = & A_{11} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_1 + A_{12} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_2 + A_{13} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_3 + \\ & + A_{21} \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_1 + A_{22} \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_2 + A_{23} \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_3 + \\ & + A_{31} \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_1 + A_{32} \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_2 + A_{33} \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_3\end{aligned}$$



# Tensor Bases – 2<sup>nd</sup> ORDER TENSOR

$$\begin{aligned}\mathbf{A} = & A_{11} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + A_{12} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2) + A_{13} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_3) + \\ & + A_{21} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + A_{22} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2) + A_{23} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_3) + \\ & + A_{31} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + A_{32} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_2) + A_{33} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_3)\end{aligned}$$

▣ In matrix notation:

$$[\mathbf{A}] = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}$$

▣ In index notation:

$$\begin{aligned}\mathbf{A} &= \sum_{ij} A_{ij} (\hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j) && \text{tensor as a physical entity} \\ [\mathbf{A}]_{ij} &= A_{ij} && \begin{array}{l} \text{component } ij \text{ of the tensor} \\ \text{in the given basis } i, j \in \{1, 2, 3\} \end{array}\end{aligned}$$

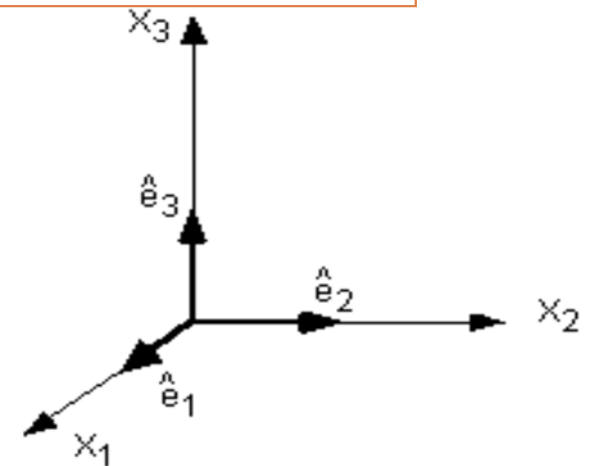
# Tensor Bases – 3<sup>rd</sup> ORDER TENSOR

- A 3<sup>rd</sup> order tensor  $\mathcal{A}$  can be written as a unique linear combination of the 27 tryads  $\hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k \equiv \hat{\mathbf{e}}_i \hat{\mathbf{e}}_j \hat{\mathbf{e}}_k$  for  $i, j, k \in \{1, 2, 3\}$ .

$$\begin{aligned} \mathcal{A} = & \mathcal{A}_{111} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{121} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{131} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{211} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{221} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{231} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{311} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{321} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{331} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{112} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2) + \mathcal{A}_{122} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2) + \dots \end{aligned}$$

Alternatively, this could have been written as:

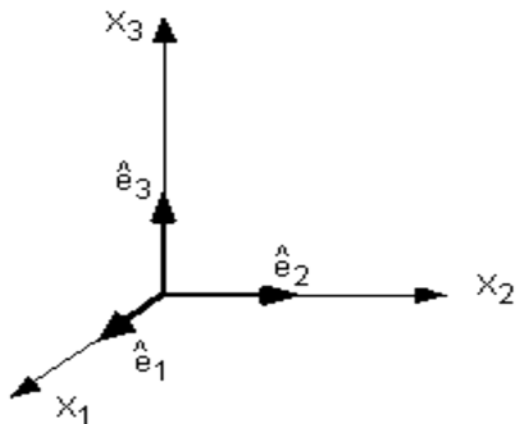
$$\begin{aligned} \mathcal{A} = & \mathcal{A}_{111} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_1 + \mathcal{A}_{121} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_1 + \mathcal{A}_{131} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_1 + \\ & + \mathcal{A}_{211} \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_1 + \mathcal{A}_{221} \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_1 + \mathcal{A}_{231} \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_1 + \\ & + \mathcal{A}_{311} \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_1 + \mathcal{A}_{321} \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_1 + \mathcal{A}_{331} \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_3 \hat{\mathbf{e}}_1 + \\ & + \mathcal{A}_{112} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_2 + \mathcal{A}_{122} \hat{\mathbf{e}}_1 \hat{\mathbf{e}}_2 \hat{\mathbf{e}}_2 + \dots \end{aligned}$$



# Tensor Bases – 3<sup>rd</sup> ORDER TENSOR

$$\begin{aligned} \mathcal{A} = & \mathcal{A}_{111} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{121} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{131} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{211} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{221} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{231} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{311} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{321} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{331} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{112} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2) + \mathcal{A}_{122} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2) + \dots \end{aligned}$$

□ In matrix notation:



$[\mathcal{A}] =$

$$\begin{bmatrix} \mathcal{A}_{111} & \mathcal{A}_{112} & \mathcal{A}_{113} & \mathcal{A}_{121} & \mathcal{A}_{122} & \mathcal{A}_{123} & \mathcal{A}_{131} & \mathcal{A}_{132} & \mathcal{A}_{133} \\ \mathcal{A}_{211} & \mathcal{A}_{212} & \mathcal{A}_{213} & \mathcal{A}_{221} & \mathcal{A}_{222} & \mathcal{A}_{223} & \mathcal{A}_{231} & \mathcal{A}_{232} & \mathcal{A}_{233} \\ \mathcal{A}_{311} & \mathcal{A}_{312} & \mathcal{A}_{313} & \mathcal{A}_{321} & \mathcal{A}_{322} & \mathcal{A}_{323} & \mathcal{A}_{331} & \mathcal{A}_{332} & \mathcal{A}_{333} \end{bmatrix}$$

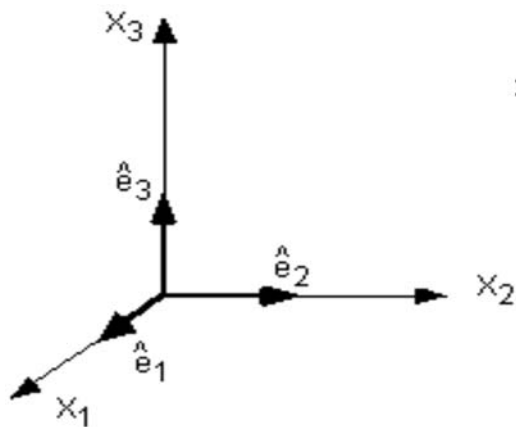
# Tensor Bases – 3<sup>rd</sup> ORDER TENSOR

$$\begin{aligned}\mathcal{A} = & \mathcal{A}_{111} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{121} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{131} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) \\ & + \mathcal{A}_{211} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{221} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{231} (\hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{311} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{321} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1) + \mathcal{A}_{331} (\hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1) + \\ & + \mathcal{A}_{112} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2) + \mathcal{A}_{122} (\hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2) + \dots\end{aligned}$$

□ In index notation:

$$\mathcal{A} = \sum_{ijk} \mathcal{A}_{ijk} (\hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k) =$$

$$= \mathcal{A}_{ijk} (\hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k) \equiv \mathcal{A}_{ijk} \hat{\mathbf{e}}_i \hat{\mathbf{e}}_j \hat{\mathbf{e}}_k \quad \text{tensor as a physical entity}$$



$$[\mathcal{A}]_{ijk} = \mathcal{A}_{ijk} \quad \text{component } ijk \text{ of the tensor in the given basis } i, j, k \in \{1, 2, 3\}$$

# Higher Order Tensors

- ▣ A tensor of order  $n$  is expressed as:

$$\mathbf{A} = A_{i_1, i_2 \dots i_n} \hat{\mathbf{e}}_{i_1} \otimes \hat{\mathbf{e}}_{i_2} \otimes \hat{\mathbf{e}}_{i_3} \otimes \dots \otimes \hat{\mathbf{e}}_{i_n}$$


where  $i_1, i_2 \dots i_n \in \{1, 2, 3\}$


- ▣ The number of components in a tensor of order  $n$  is  $3^n$ .



# Repeated-index (or Einstein's) Notation

- The **Einstein Summation Convention**: repeated Roman indices are summed over.

$i$  is a mute index   $a_i b_i = \sum_{i=1}^3 a_i b_i = a_1 b_1 + a_2 b_2 + a_3 b_3$

$i$  is a talking index and  $j$  is a mute index   $A_{ij} b_j = \sum_{j=1}^3 A_{ij} b_j = A_{i1} b_1 + A_{i2} b_2 + A_{i3} b_3$

- A **“MUTE” (or DUMMY) INDEX** is an index that does not appear in a monomial after the summation is carried out (it can be arbitrarily changed of “name”).
- A **“TALKING” INDEX** is an index that is not repeated in the same monomial and is transmitted outside of it (it cannot be arbitrarily changed of “name”).

## REMARK

An index can only appear up to two times in a monomial.

# Repeated-index (or Einstein's) Notation

## Rules of this notation:

1. Sum over all repeated indices.
2. Increment all unique indices fully at least once, covering all combinations.
3. Increment repeated indices first.

4. A comma indicates differentiation, with respect to coordinate  $x_i$ .

$$u_{i,i} = \frac{\partial u_i}{\partial x_i} = \sum_{i=1}^3 \frac{\partial u_i}{\partial x_i} \quad u_{i,jj} = \frac{\partial^2 u_i}{\partial x_j \partial x_j} = \sum_{j=1}^3 \frac{\partial^2 u_i}{\partial x_j^2} \quad A_{ij,j} = \frac{\partial A_{ij}}{\partial x_j} = \sum_{j=1}^3 \frac{\partial A_{ij}}{\partial x_j}$$

5. The number of talking indices indicates the order of the tensor result

# Kronecker Delta $\delta$

- ▣ The Kronecker delta  $\delta_{ij}$  is defined as:

$$\delta_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

- Both  $i$  and  $j$  may take on any value in  $\{1, 2, 3\}$
- Only for the three possible cases where  $i = j$  is  $\delta_{ij}$  non-zero.

$$\delta_{ij} = \begin{cases} 1 & \text{if } i = j \quad (\delta_{11} = \delta_{22} = \delta_{33} = 1) \\ 0 & \text{if } i \neq j \quad (\delta_{12} = \delta_{13} = \delta_{21} \dots = 0) \end{cases}$$

- $\delta_{ij} = \delta_{ji}$

## REMARK

Following Einstein's notation:  $\delta_{ii} = \delta_{11} + \delta_{22} + \delta_{33} = 3$   
Kronecker delta serves as a replacement operator:

$$\delta_{ij} u_j = u_i \quad , \quad \delta_{ij} A_{jk} = A_{ik}$$

# Levi-Civita Epsilon (permutation) $\epsilon$

- ▣ The Levi-Civita epsilon  $\epsilon_{ijk}$  is defined as:

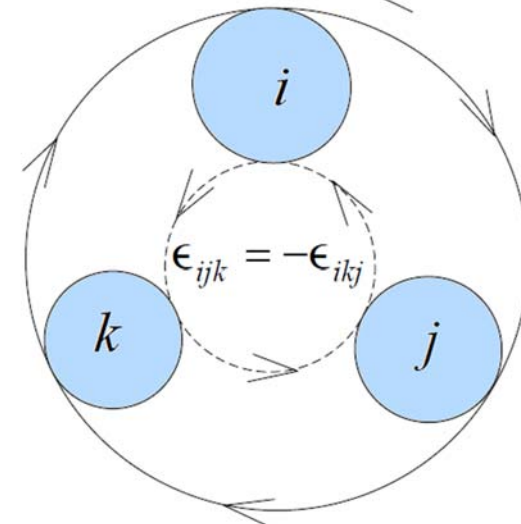
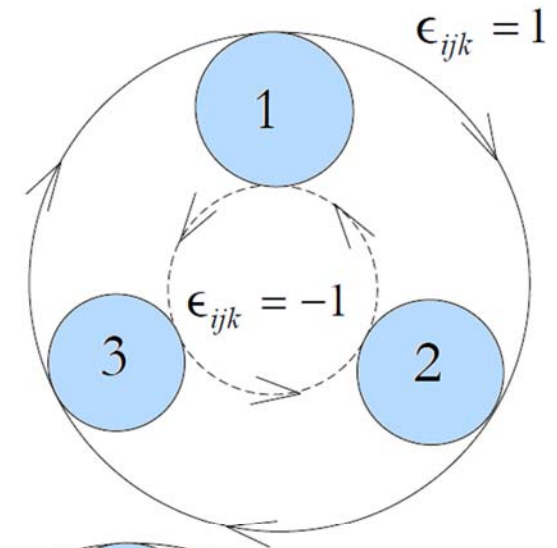
$$\epsilon_{ijk} = \begin{cases} 0 & \text{if there is a repeated index} \\ +1 & \text{if } ijk = 123, 231 \text{ or } 312 \\ -1 & \text{if } ijk = 213, 132 \text{ or } 321 \end{cases}$$

- 3 indices  $\Rightarrow$  27 possible combinations.

- $\epsilon_{ijk} = -\epsilon_{ikj}$

## REMARK

The Levi-Civita symbol is also named permutation or alternating symbol.



# Relation between $\delta$ and $\varepsilon$

$$\epsilon_{ijk} = \det \begin{bmatrix} \delta_{i1} & \delta_{i2} & \delta_{i3} \\ \delta_{j1} & \delta_{j2} & \delta_{j3} \\ \delta_{k1} & \delta_{k2} & \delta_{k3} \end{bmatrix}$$

$$\epsilon_{ijk} \epsilon_{pqr} = \det \begin{bmatrix} \delta_{ip} & \delta_{iq} & \delta_{ir} \\ \delta_{jp} & \delta_{jq} & \delta_{jr} \\ \delta_{kp} & \delta_{kq} & \delta_{kr} \end{bmatrix}$$

$$\epsilon_{ijk} \epsilon_{pqk} = \delta_{ip} \delta_{jq} - \delta_{iq} \delta_{jp}$$

$$\epsilon_{ijk} \epsilon_{pjk} = 2\delta_{pi}$$

$$\epsilon_{ijk} \epsilon_{ijk} = 6$$

# Example

- Prove the following expression is true:

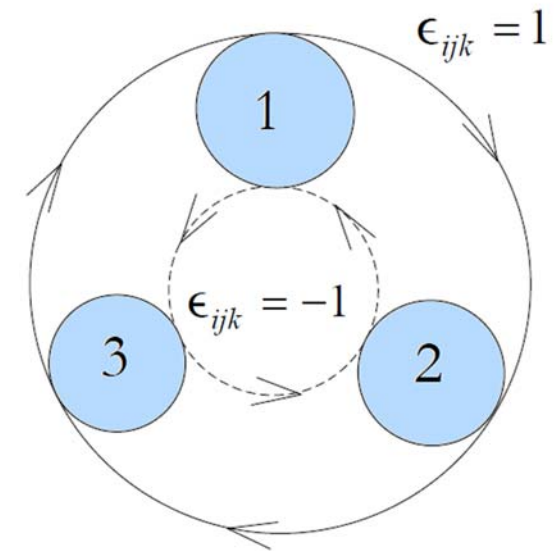
$$\epsilon_{ijk} \epsilon_{ijk} = 6$$

# Example - Solution

$$\begin{aligned}
 \epsilon_{ijk} \epsilon_{ijk} &= \epsilon_{111} \epsilon_{111} + \epsilon_{112} \epsilon_{112} + \epsilon_{113} \epsilon_{113} + \epsilon_{121} \epsilon_{121} + \epsilon_{122} \epsilon_{122} + \epsilon_{123} \epsilon_{123} + \epsilon_{131} \epsilon_{131} + \epsilon_{132} \epsilon_{132} + \epsilon_{133} \epsilon_{133} + \epsilon_{211} \epsilon_{211} + \epsilon_{212} \epsilon_{212} + \epsilon_{213} \epsilon_{213} + \epsilon_{221} \epsilon_{221} + \epsilon_{222} \epsilon_{222} + \epsilon_{223} \epsilon_{223} + \epsilon_{231} \epsilon_{231} + \epsilon_{232} \epsilon_{232} + \epsilon_{233} \epsilon_{233} + \epsilon_{311} \epsilon_{311} + \epsilon_{312} \epsilon_{312} + \epsilon_{313} \epsilon_{313} + \epsilon_{321} \epsilon_{321} + \epsilon_{322} \epsilon_{322} + \epsilon_{323} \epsilon_{323} + \epsilon_{331} \epsilon_{331} + \epsilon_{332} \epsilon_{332} + \epsilon_{333} \epsilon_{333} = 6
 \end{aligned}$$

$k=1$     $k=2$     $k=3$   
 $i=1$     $j=1$   
 $j=2$   
 $j=3$   
 $i=2$   
 $i=3$

(Circled terms in the original image indicate the values of  $\epsilon_{ijk}$  for each  $i, j, k$  combination:  $\epsilon_{123} = 1$ ,  $\epsilon_{132} = -1$ ,  $\epsilon_{213} = -1$ ,  $\epsilon_{231} = 1$ ,  $\epsilon_{312} = 1$ ,  $\epsilon_{321} = -1$ . All other terms are zero.)





# Vector Operations

Tensor Algebra

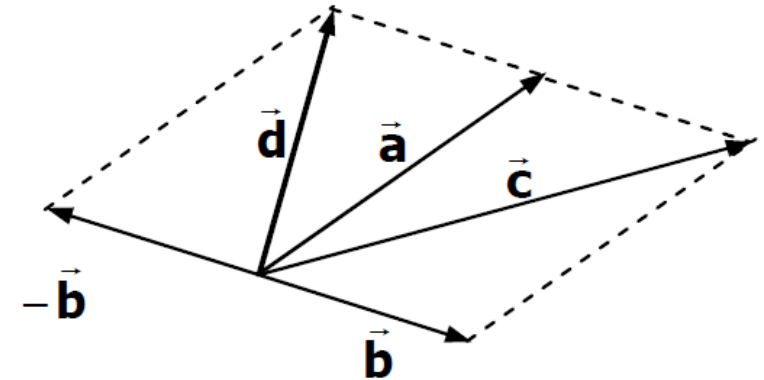


# Vector Operations

## ▣ Sum and Subtraction. Parallelogram law.

$$\mathbf{a} + \mathbf{b} = \mathbf{b} + \mathbf{a} = \mathbf{c} \quad \Rightarrow \quad c_i = a_i + b_i$$

$$\mathbf{a} - \mathbf{b} = \mathbf{d} \quad \Rightarrow \quad d_i = a_i - b_i$$



## ▣ Scalar multiplication

$$\alpha \mathbf{a} = \mathbf{b} = \alpha a_1 \hat{\mathbf{e}}_1 + \alpha a_2 \hat{\mathbf{e}}_2 + \alpha a_3 \hat{\mathbf{e}}_3 \quad \Rightarrow \quad b_i = \alpha a_i$$

# Vector Operations

- ▣ **Scalar or dot product** yields a scalar

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta$$

where  $\theta$  is the angle between the vectors  $\mathbf{u}$  and  $\mathbf{v}$

- In index notation:

$$\mathbf{u} \cdot \mathbf{v} = u_i \underbrace{\hat{\mathbf{e}}_i}_{\mathbf{u}} \cdot v_j \underbrace{\hat{\mathbf{e}}_j}_{\mathbf{v}} = u_i v_j \underbrace{\hat{\mathbf{e}}_i \cdot \hat{\mathbf{e}}_j}_{\delta_{ij}} = u_i v_j \delta_{ij} = \underbrace{u_i v_i}_{\substack{=0 (i \neq j) \\ =1 (j=i)}} \left( = \sum_{i=1}^{i=3} u_i v_i \right) = [\mathbf{u}]^T [\mathbf{v}]$$

- ▣ **Norm of a vector**

$$\|\mathbf{u}\| = (\mathbf{u} \cdot \mathbf{u})^{1/2} = (u_i u_i)^{1/2}$$

$$\Rightarrow \|\mathbf{u}\|^2 = \mathbf{u} \cdot \mathbf{u} = u_i \hat{\mathbf{e}}_i \cdot u_j \hat{\mathbf{e}}_j = u_i u_j \delta_{ij} = u_i u_i$$

# Vector Operations

- ▣ Some **properties** of the **scalar or dot product**

$$\mathbf{u} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{u}$$

$$\mathbf{u} \cdot \mathbf{0} = 0$$

$$\mathbf{u} \cdot (\alpha \mathbf{v} + \beta \mathbf{w}) = \alpha (\mathbf{u} \cdot \mathbf{v}) + \beta (\mathbf{u} \cdot \mathbf{w}) \Rightarrow \text{Linear operator}$$

$$\mathbf{u} \cdot \mathbf{u} > 0 \iff \mathbf{u} \neq \mathbf{0}$$

$$\mathbf{u} \cdot \mathbf{u} = 0 \iff \mathbf{u} = \mathbf{0}$$

$$\mathbf{u} \cdot \mathbf{v} = 0, \quad \mathbf{u} \neq \mathbf{0}, \quad \mathbf{v} \neq \mathbf{0} \iff \mathbf{u} \perp \mathbf{v}$$

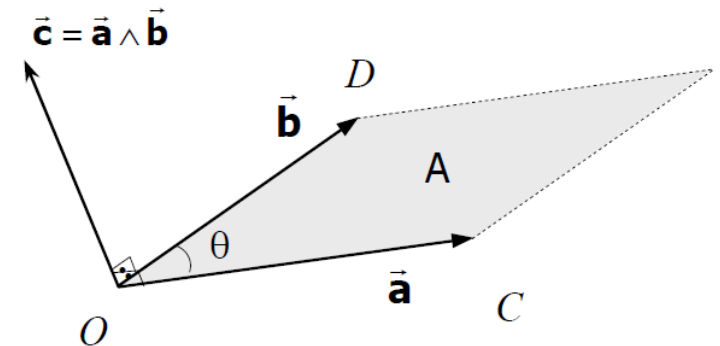
# Vector Operations

▣ **Vector product (or cross product)** yields another vector

$$\mathbf{c} = \mathbf{a} \times \mathbf{b} = -\mathbf{b} \times \mathbf{a}$$

$$|\mathbf{c}| = |\mathbf{a}||\mathbf{b}|\sin\theta \quad \text{where } \theta \text{ is the angle between the vectors } \mathbf{a} \text{ and } \mathbf{b}$$

$$0 \leq \theta \leq \pi$$



■ In index notation:

$$\mathbf{c} = c_i \hat{\mathbf{e}}_i = \epsilon_{ijk} a_j b_k \hat{\mathbf{e}}_i \Rightarrow c_i = \epsilon_{ijk} a_j b_k \quad i \in \{1, 2, 3\}$$

$$\begin{aligned} \mathbf{c} = & \underbrace{(a_2 b_3 - a_3 b_2)}_{\substack{\epsilon_{123} a_2 b_3 + \epsilon_{132} a_3 b_2 \\ \epsilon_{123} = 1, \epsilon_{132} = -1}} \hat{\mathbf{e}}_1 + \underbrace{(a_3 b_1 - a_1 b_3)}_{\substack{\epsilon_{231} a_3 b_1 + \epsilon_{213} a_1 b_3 \\ \epsilon_{231} = 1, \epsilon_{213} = -1}} \hat{\mathbf{e}}_2 + \underbrace{(a_1 b_2 - a_2 b_1)}_{\substack{\epsilon_{312} a_1 b_2 + \epsilon_{321} a_2 b_1 \\ \epsilon_{312} = 1, \epsilon_{321} = -1}} \hat{\mathbf{e}}_3 \end{aligned} \quad \overset{\text{symp}}{=} \det \begin{bmatrix} \hat{\mathbf{e}}_1 & \hat{\mathbf{e}}_2 & \hat{\mathbf{e}}_3 \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{bmatrix}$$

# Vector Operations

- Some **properties** of the **vector or cross product**

$$\mathbf{u} \times \mathbf{v} = -(\mathbf{v} \times \mathbf{u})$$

$$\mathbf{u} \times \mathbf{v} = \mathbf{0}, \quad \mathbf{u} \neq \mathbf{0}, \mathbf{v} \neq \mathbf{0} \quad \longleftrightarrow \quad \mathbf{u} \parallel \mathbf{v}$$

$$\mathbf{u} \times (a\mathbf{v} + b\mathbf{w}) = a\mathbf{u} \times \mathbf{v} + b\mathbf{u} \times \mathbf{w} \quad \Rightarrow \quad \text{Linear operator}$$

# Vector Operations

- ▣ **Tensor product (or open or dyadic product) of two vectors:**

$$\mathbf{A} = \mathbf{u} \otimes \mathbf{v} \equiv \mathbf{u}\mathbf{v}$$

Also known as the **dyad** of the vectors  $\mathbf{u}$  and  $\mathbf{v}$ , which results in a 2<sup>nd</sup> order tensor  $\mathbf{A}$ .

- Deriving the tensor product along an orthonormal basis  $\{\hat{\mathbf{e}}_i\}$ :

$$\mathbf{A} = (\mathbf{u} \otimes \mathbf{v}) = (u_i \hat{\mathbf{e}}_i) \otimes (v_j \hat{\mathbf{e}}_j) = u_i v_j (\hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j) = A_{ij} (\hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j)$$

$$[\mathbf{A}]_{ij} = A_{ij} = [\mathbf{u} \otimes \mathbf{v}]_{ij} = u_i v_j \quad i, j \in \{1, 2, 3\}$$

- In matrix notation:

$$[\mathbf{u} \otimes \mathbf{v}] = [\mathbf{u}][\mathbf{v}]^T = \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} \begin{bmatrix} v_1 & v_2 & v_3 \end{bmatrix} = \begin{bmatrix} u_1 v_1 & u_1 v_2 & u_1 v_3 \\ u_2 v_1 & u_2 v_2 & u_2 v_3 \\ u_3 v_1 & u_3 v_2 & u_3 v_3 \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}$$

# Vector Operations

## ▣ Some **properties** of the **open product**:

$$(\mathbf{u} \otimes \mathbf{v}) \neq (\mathbf{v} \otimes \mathbf{u})$$

$$(\mathbf{u} \otimes \mathbf{v}) \cdot \mathbf{w} = \mathbf{u} \otimes (\mathbf{v} \cdot \mathbf{w}) = \mathbf{u} (\mathbf{v} \cdot \mathbf{w}) = (\mathbf{v} \cdot \mathbf{w}) \mathbf{u}$$

$$\mathbf{u} \otimes (\alpha \mathbf{v} + \beta \mathbf{w}) = \alpha \mathbf{u} \otimes \mathbf{v} + \beta \mathbf{u} \otimes \mathbf{w} \quad \Rightarrow \quad \text{Linear operator}$$

$$(\mathbf{u} \otimes \mathbf{v})(\mathbf{w} \otimes \mathbf{x}) = (\mathbf{u} \otimes \mathbf{x})(\mathbf{v} \cdot \mathbf{w})$$

$$\mathbf{u} \cdot (\mathbf{v} \otimes \mathbf{w}) = (\mathbf{u} \cdot \mathbf{v}) \otimes \mathbf{w} = (\mathbf{u} \cdot \mathbf{v}) \mathbf{w} = \mathbf{w} (\mathbf{u} \cdot \mathbf{v})$$

# Example

- Prove the following property of the tensor product is true:

$$\mathbf{u} \cdot (\mathbf{v} \otimes \mathbf{w}) = (\mathbf{u} \cdot \mathbf{v}) \otimes \mathbf{w}$$



# Example - Solution

$$\underbrace{\mathbf{u}}_{\substack{\text{vector} \\ \text{1}^{\text{st}} \text{ order tensor} \\ \text{(vector)}}} \cdot \underbrace{(\mathbf{v} \otimes \mathbf{w})}_{\substack{\text{2}^{\text{nd}} \text{ order} \\ \text{tensor (matrix)}}} = \underbrace{(\mathbf{u} \cdot \mathbf{v})}_{\substack{\text{scalar} \\ \text{1}^{\text{st}} \text{ order tensor} \\ \text{(vector)}}} \otimes \underbrace{\mathbf{w}}_{\text{vector}} \Rightarrow \mathbf{c}$$

$$c_k = [\mathbf{u} \cdot (\mathbf{v} \otimes \mathbf{w})]_k = [\mathbf{u}]_i [(\mathbf{v} \otimes \mathbf{w})]_{ik} = u_i (v_i w_k) = u_i v_i w_k \Rightarrow k\text{-component of vector } \mathbf{c}$$

$$c_k = [(\mathbf{u} \cdot \mathbf{v}) \otimes \mathbf{w}]_k = u_i v_i \otimes [\mathbf{w}]_k = u_i v_i w_k \Rightarrow k\text{-component of vector } \mathbf{c}$$

$$\mathbf{c} = u_i v_j w_k \hat{\mathbf{e}}_k = [\mathbf{u} \cdot (\mathbf{v} \otimes \mathbf{w})]_k \hat{\mathbf{e}}_k = [(\mathbf{u} \cdot \mathbf{v}) \otimes \mathbf{w}]_k \hat{\mathbf{e}}_k$$

# Example – Solution

$$\mathbf{u} \cdot (\mathbf{v} \otimes \mathbf{w}) = (\mathbf{u} \cdot \mathbf{v}) \otimes \mathbf{w}$$

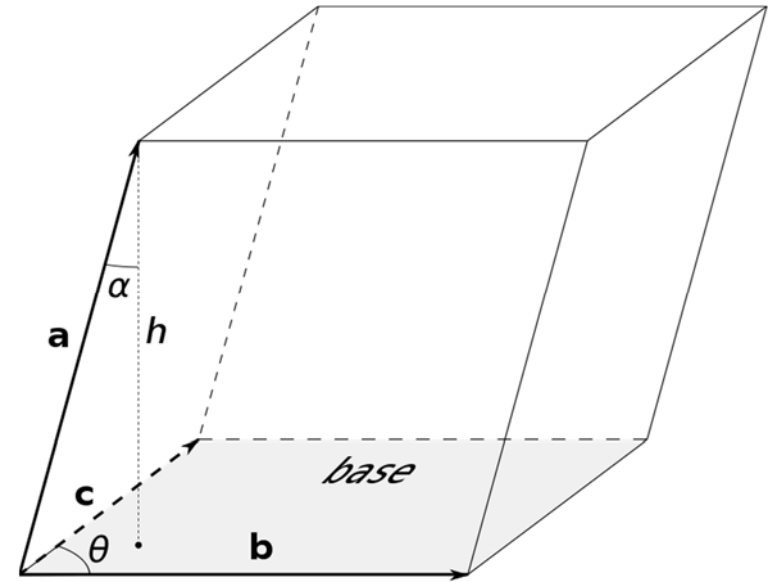
$$\mathbf{u} \cdot (\mathbf{v} \otimes \mathbf{w}) = u_i \hat{\mathbf{e}}_i \cdot (\mathbf{v}_j \hat{\mathbf{e}}_j \otimes \mathbf{w}_k \hat{\mathbf{e}}_k) = u_i \hat{\mathbf{e}}_i \cdot (\mathbf{v}_j \mathbf{w}_k \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k) = u_i \mathbf{v}_j \mathbf{w}_k (\hat{\mathbf{e}}_i \cdot \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k)$$

$$(\mathbf{u} \cdot \mathbf{v}) \otimes \mathbf{w} = (u_i \hat{\mathbf{e}}_i \cdot \mathbf{v}_j \hat{\mathbf{e}}_j) \otimes \mathbf{w}_k \hat{\mathbf{e}}_k = (u_i \mathbf{v}_j \hat{\mathbf{e}}_i \cdot \hat{\mathbf{e}}_j) \otimes \mathbf{w}_k \hat{\mathbf{e}}_k = u_i \mathbf{v}_j \mathbf{w}_k (\hat{\mathbf{e}}_i \cdot \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k)$$

# Vector Operations

## ▣ Triple scalar or box product

$$\begin{aligned} V &= \mathbf{a} \cdot (\mathbf{b} \times \mathbf{c}) \\ &= \mathbf{c} \cdot (\mathbf{a} \times \mathbf{b}) = \mathbf{b} \cdot (\mathbf{c} \times \mathbf{a}) \\ &= |\mathbf{a}| \cos \alpha |\mathbf{b} \times \mathbf{c}| \\ &= \underbrace{|\mathbf{a}| \cos \alpha}_{\text{height}} \underbrace{|\mathbf{b}| |\mathbf{c}| \sin \theta}_{\text{base area}} \end{aligned}$$



■ In index notation:

$$V = \mathbf{a} \cdot (\mathbf{b} \times \mathbf{c}) = \epsilon_{ijk} a_i b_j c_k \quad \Rightarrow \quad V = \mathbf{a} \cdot (\mathbf{b} \times \mathbf{c}) = \det \begin{bmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{bmatrix}$$

# Vector Operations

## ▣ Triple vector product

$$\mathbf{u} \times (\mathbf{v} \times \mathbf{w}) = (\mathbf{u} \cdot \mathbf{w}) \mathbf{v} - (\mathbf{u} \cdot \mathbf{v}) \mathbf{w}$$

■ In index notation:

$$\begin{aligned}\mathbf{u} \times (\mathbf{v} \times \mathbf{w}) &= u_j \hat{\mathbf{e}}_j \times (\epsilon_{klm} v_l w_m \hat{\mathbf{e}}_k) = \epsilon_{ijk} u_j (\epsilon_{klm} v_l w_m) \hat{\mathbf{e}}_i = \\ &= \epsilon_{ijk} \epsilon_{lmk} u_j v_l w_m \hat{\mathbf{e}}_i = (\delta_{il} \delta_{jm} - \delta_{im} \delta_{jl}) u_j v_l w_m \hat{\mathbf{e}}_i = \\ &= u_m v_i w_m \hat{\mathbf{e}}_i - u_l v_l w_i \hat{\mathbf{e}}_i\end{aligned}$$

### REMARK

$$\epsilon_{ijk} \epsilon_{pqk} = \delta_{ip} \delta_{jq} - \delta_{iq} \delta_{jp}$$

# Summary

- ▣ **Scalar or dot product** yields a scalar

$$\mathbf{u} \cdot \mathbf{v} = [\mathbf{u}]^T [\mathbf{v}] = |\mathbf{u}| |\mathbf{v}| \cos \theta$$

$$\mathbf{u} \cdot \mathbf{v} = u_i v_i$$

- ▣ **Vector or cross product** yields another vector

$$\mathbf{c} = \mathbf{a} \times \mathbf{b} = -\mathbf{b} \times \mathbf{a}$$

$$[\mathbf{a} \times \mathbf{b}]_i = c_i = \epsilon_{ijk} a_j b_k$$

- ▣ **Triple scalar or box product** yields a scalar

$$V = \mathbf{a} \cdot (\mathbf{b} \times \mathbf{c}) = |\mathbf{a}| \cos \alpha |\mathbf{b}| |\mathbf{c}| \sin \theta$$

$$\mathbf{a} \cdot (\mathbf{b} \times \mathbf{c}) = \epsilon_{ijk} a_i b_j c_k$$

- ▣ **Triple vector product** yields another vector

$$\mathbf{u} \times (\mathbf{v} \times \mathbf{w}) = (\mathbf{u} \cdot \mathbf{w}) \mathbf{v} - (\mathbf{u} \cdot \mathbf{v}) \mathbf{w}$$

$$[\mathbf{u} \times (\mathbf{v} \times \mathbf{w})]_i = u_k w_k v_i - u_k v_k w_i$$



# Tensor Operations

## Tensor Algebra

# Tensor Operations

- ▣ **Summation (only for equal order tensors)**

$$\mathbf{A} + \mathbf{B} = \mathbf{B} + \mathbf{A} = \mathbf{C} \quad \Rightarrow \quad C_{ij} = A_{ij} + B_{ij}$$

- ▣ **Scalar multiplication (scalar times tensor)**

$$\alpha \mathbf{A} = \mathbf{C} \quad \Rightarrow \quad C_{ij} = \alpha A_{ij}$$

# Tensor Operations

## □ Dot product (.) or single index contraction product

$$\underbrace{\mathbf{A}}_{\substack{2^{\text{nd}} \\ \text{order}}} \cdot \underbrace{\mathbf{b}}_{\substack{1^{\text{st}} \\ \text{order}}} = \underbrace{\mathbf{c}}_{\substack{1^{\text{st}} \\ \text{order}}} \quad \Rightarrow \quad c_i = A_{ij} b_j \quad \Rightarrow \quad \text{Index "j" disappears (index contraction)}$$

$$\underbrace{\mathcal{A}}_{\substack{3^{\text{rd}} \\ \text{order}}} \cdot \underbrace{\mathbf{b}}_{\substack{1^{\text{st}} \\ \text{order}}} = \underbrace{\mathbf{C}}_{\substack{2^{\text{nd}} \\ \text{order}}} \quad \Rightarrow \quad C_{ij} = \mathcal{A}_{ijk} b_k \quad \Rightarrow \quad \text{Index "k" disappears (index contraction)}$$

$$\underbrace{\mathbf{A}}_{\substack{2^{\text{nd}} \\ \text{order}}} \cdot \underbrace{\mathbf{B}}_{\substack{2^{\text{nd}} \\ \text{order}}} = \underbrace{\mathbf{C}}_{\substack{2^{\text{nd}} \\ \text{order}}} \quad \Rightarrow \quad C_{ik} = A_{ij} B_{jk} \quad \Rightarrow \quad \text{Index "j" disappears (index contraction)}$$

$$\mathbf{A} \cdot \mathbf{B} \neq \mathbf{B} \cdot \mathbf{A}$$

**REMARK**

$$\mathbf{A} \cdot \mathbf{A} = \mathbf{A}^2$$



# Tensor Operations

## ▣ Some **properties**:

$$\mathbf{A} \cdot (\alpha \mathbf{b} + \beta \mathbf{c}) = \alpha \mathbf{A} \cdot \mathbf{b} + \beta \mathbf{A} \cdot \mathbf{c} \quad \Rightarrow \quad \text{Linear operator}$$

## ▣ **2<sup>nd</sup> order unit (or identity) tensor**

$$\mathbf{1} \cdot \mathbf{u} = \mathbf{u} \cdot \mathbf{1} = \mathbf{u}$$

$$\begin{cases} \mathbf{1} = \delta_{ij} \mathbf{e}_j \otimes \mathbf{e}_i = \mathbf{e}_i \otimes \mathbf{e}_i \\ [1]_{ij} = \delta_{ij} \end{cases}$$

$$[\mathbf{1}] = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

# 2<sup>nd</sup> Order Tensor Operations

▣ Some **properties**:

$$\mathbf{1} \cdot \mathbf{A} = \mathbf{A} = \mathbf{A} \cdot \mathbf{1}$$

$$\mathbf{A} \cdot (\mathbf{B} + \mathbf{C}) = \mathbf{A} \cdot \mathbf{B} + \mathbf{A} \cdot \mathbf{C}$$

$$\mathbf{A} \cdot (\mathbf{B} \cdot \mathbf{C}) = (\mathbf{A} \cdot \mathbf{B}) \cdot \mathbf{C} = \mathbf{A} \cdot \mathbf{B} \cdot \mathbf{C}$$

$$\mathbf{A} \cdot \mathbf{B} \neq \mathbf{B} \cdot \mathbf{A}$$

# Example

- When does the relation  $\mathbf{n} \cdot \mathbf{T} = \mathbf{T} \cdot \mathbf{n}$  hold true ?

# Example - Solution

$$\underbrace{\underbrace{\mathbf{n}}_{\text{vector}} \cdot \underbrace{\mathbf{T}}_{\text{2nd order tensor}}}_{\text{vector}} = \mathbf{T} \cdot \mathbf{n} \Rightarrow \mathbf{c}$$

$$\mathbf{c} = \mathbf{n} \cdot \mathbf{T} \Rightarrow c_k = n_i T_{ik}$$

$$\mathbf{c}^* = \mathbf{T} \cdot \mathbf{n} \Rightarrow c_k^* = T_{ki} n_i = n_i T_{ki}$$

$$c_k = c_k^* \quad \text{if} \quad T_{ik} = T_{ki}$$

$$\mathbf{c} = c_k \hat{\mathbf{e}}_k = [\mathbf{n} \cdot \mathbf{T}]_k \hat{\mathbf{e}}_k = n_i T_{ik} \hat{\mathbf{e}}_k$$

$$\mathbf{c}^* = c_k^* \hat{\mathbf{e}}_k = [\mathbf{T} \cdot \mathbf{n}]_k \hat{\mathbf{e}}_k = n_i T_{ki} \hat{\mathbf{e}}_k$$

# Example - Solution

$$\mathbf{n} \cdot \mathbf{T} = \mathbf{T} \cdot \mathbf{n} \quad \Rightarrow \quad \text{COMPACT NOTATION}$$

$$n_i T_{ik} = T_{ki} n_i \quad k \in \{1, 2, 3\} \quad \Rightarrow \quad \text{INDEX NOTATION}$$

$$\underbrace{\underbrace{[\mathbf{n}]^T \ [\mathbf{T}]}_{\mathbf{c}^T}}_{\mathbf{c}} = \underbrace{[\mathbf{T}][\mathbf{n}]}_{\mathbf{c}} \quad \Rightarrow \quad \text{MATRIX NOTATION}$$

# Example - Solution

$$\underbrace{\underbrace{[\mathbf{n}]^T [\mathbf{T}]}_{\mathbf{c}^T}}_{\mathbf{c}} = \underbrace{[\mathbf{T}]}_{\mathbf{c}} [\mathbf{n}] \quad \Rightarrow \quad \text{MATRIX NOTATION}$$

$$\begin{bmatrix} n_1 & n_2 & n_3 \end{bmatrix} \begin{bmatrix} T_{11} & T_{12} & T_{13} \\ T_{21} & T_{22} & T_{23} \\ T_{31} & T_{32} & T_{33} \end{bmatrix}^T = \begin{bmatrix} T_{11} & T_{12} & T_{13} \\ T_{21} & T_{22} & T_{23} \\ T_{31} & T_{32} & T_{33} \end{bmatrix} \begin{bmatrix} n_1 \\ n_2 \\ n_3 \end{bmatrix}$$

$$\begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix}^T = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}$$

# 2<sup>nd</sup> Order Tensor Operations

## ▣ Transpose

$$[\mathbf{A}] = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix} \Rightarrow [\mathbf{A}^T] = \begin{bmatrix} A_{11} & A_{21} & A_{31} \\ A_{12} & A_{22} & A_{32} \\ A_{13} & A_{23} & A_{33} \end{bmatrix} \left\{ \begin{array}{l} (\mathbf{A}^T)^T = \mathbf{A} \\ (\mathbf{A} \cdot \mathbf{B})^T = \mathbf{B}^T \cdot \mathbf{A}^T \\ (\mathbf{u} \otimes \mathbf{v})^T = \mathbf{v} \otimes \mathbf{u} \\ (\alpha \mathbf{A} + \beta \mathbf{B})^T = \alpha \mathbf{A}^T + \beta \mathbf{B}^T \end{array} \right.$$
$$[\mathbf{A}^T]_{ij} = A_{ji}$$

## ▣ Trace yields a scalar

$$\text{Tr}(\mathbf{A}) = A_{ii} (= A_{11} + A_{22} + A_{33})$$

$$\text{Tr}(\mathbf{a} \otimes \mathbf{b}) = \text{Tr}[a_i b_j] = a_i b_i = \mathbf{a} \cdot \mathbf{b}$$

## ■ Some properties:

$$\text{Tr}(\mathbf{A}^T) = \text{Tr} \mathbf{A}$$

$$\text{Tr}(\mathbf{A} + \mathbf{B}) = \text{Tr} \mathbf{A} + \text{Tr} \mathbf{B}$$

$$\text{Tr}(\alpha \mathbf{A}) = \alpha \text{Tr} \mathbf{A}$$

$$\text{Tr}(\mathbf{A} \cdot \mathbf{B}) = \text{Tr}(\mathbf{B} \cdot \mathbf{A})$$

# 2<sup>nd</sup> Order Tensor Operations

## □ Double index contraction or **double (vertical) dot product** (:)

$$\underbrace{\mathbf{A}}_{\substack{2^{\text{nd}} \\ \text{order}}} : \underbrace{\mathbf{B}}_{\substack{2^{\text{nd}} \\ \text{order}}} = \underbrace{c}_{\substack{\text{zero} \\ \text{order} \\ \text{(scalar)}}} \Rightarrow c = A_{ij} B_{ij} \Rightarrow \text{Indices "i,j" disappear (double index contraction)}$$

$$\underbrace{\mathcal{A}}_{\substack{3^{\text{rd}} \\ \text{order}}} : \underbrace{\mathbf{B}}_{\substack{2^{\text{nd}} \\ \text{order}}} = \underbrace{\mathbf{c}}_{\substack{1^{\text{st}} \\ \text{order}}} \Rightarrow \mathbf{c}_i = \mathcal{A}_{ijk} B_{jk} \Rightarrow \text{Indices "j,k" disappear (double index contraction)}$$

$$\underbrace{\mathbb{A}}_{\substack{4^{\text{th}} \\ \text{order}}} : \underbrace{\mathbf{B}}_{\substack{2^{\text{nd}} \\ \text{order}}} = \underbrace{\mathbf{C}}_{\substack{2^{\text{nd}} \\ \text{order}}} \Rightarrow C_{ij} = A_{ijkl} B_{kl} \Rightarrow \text{Indices "k,l" disappear (double index contraction)}$$

- **Indices contiguous to the double-dot (:) operator get vertically repeated (contraction) and they disappear in the resulting tensor (4 order reduction of the sum of orders).**



# 2<sup>nd</sup> Order Tensor Operations

## ▣ Some properties

$$\mathbf{A} : \mathbf{B} = \text{Tr}(\mathbf{A}^T \cdot \mathbf{B}) = \text{Tr}(\mathbf{B}^T \cdot \mathbf{A}) = \text{Tr}(\mathbf{A} \cdot \mathbf{B}^T) = \text{Tr}(\mathbf{B} \cdot \mathbf{A}^T) = \mathbf{B} : \mathbf{A}$$

$$\mathbf{1} : \mathbf{A} = \text{Tr} \mathbf{A} = \mathbf{A} : \mathbf{1}$$

$$\mathbf{A} : (\mathbf{B} \cdot \mathbf{C}) = (\mathbf{B}^T \cdot \mathbf{A}) : \mathbf{C} = (\mathbf{A} \cdot \mathbf{C}^T) : \mathbf{B}$$

$$\mathbf{A} : (\mathbf{u} \otimes \mathbf{v}) = \mathbf{u} \cdot (\mathbf{A} \cdot \mathbf{v})$$

$$(\mathbf{u} \otimes \mathbf{v}) : (\mathbf{w} \otimes \mathbf{x}) = (\mathbf{u} \cdot \mathbf{w}) \cdot (\mathbf{v} \cdot \mathbf{x})$$

### REMARK

$$\mathbf{A} : \mathbf{B} = \mathbf{C} : \mathbf{B} \not\Rightarrow \mathbf{A} = \mathbf{C}$$

# 2<sup>nd</sup> Order Tensor Operations

## □ Double index contraction or **double (horizontal) dot product** (··)

$$\underbrace{\mathbf{A}}_{\substack{2^{\text{nd}} \\ \text{order}}} \cdot \cdot \underbrace{\mathbf{B}}_{\substack{2^{\text{nd}} \\ \text{order}}} = \underbrace{c}_{\substack{\text{zero} \\ \text{order} \\ \text{(scalar)}}} \Rightarrow c = A_{ij} B_{ji} \Rightarrow \text{Indices "i,j" disappear (double index contraction)}$$

$$\underbrace{\mathcal{A}}_{\substack{3^{\text{rd}} \\ \text{order}}} \cdot \cdot \underbrace{\mathbf{B}}_{\substack{2^{\text{nd}} \\ \text{order}}} = \underbrace{\mathbf{c}}_{\substack{1^{\text{st}} \\ \text{order}}} \Rightarrow \mathbf{c}_i = \mathcal{A}_{ijk} B_{kj} \Rightarrow \text{Indices "j,k" disappear (double index contraction)}$$

$$\underbrace{\mathbb{A}}_{\substack{4^{\text{th}} \\ \text{order}}} \cdot \cdot \underbrace{\mathbf{B}}_{\substack{2^{\text{nd}} \\ \text{order}}} = \underbrace{\mathbf{C}}_{\substack{2^{\text{nd}} \\ \text{order}}} \Rightarrow C_{ij} = \mathbb{A}_{ijkl} B_{lk} \Rightarrow \text{Indices "k,l" disappear (double index contraction)}$$

- **Indices contiguous to the double-dot (··) operator** get horizontally repeated (contraction) and they **disappear** in the resulting tensor (4 orders reduction of the sum of orders).

# Tensor Operations

- ▣ **Norm** of a tensor is a non-negative real number defined by

$$\|\mathbf{A}\| = (\mathbf{A} : \mathbf{A})^{1/2} = (A_{ij} A_{ij})^{1/2} \geq 0$$

$$\mathbf{A} \cdot \cdot \mathbf{B} = \text{Tr}(\mathbf{A} \cdot \mathbf{B}) = \text{Tr}(\mathbf{B} \cdot \mathbf{A}) = \mathbf{B} \cdot \cdot \mathbf{A}$$

$$\mathbf{1} \cdot \cdot \mathbf{A} = \text{Tr} \mathbf{A} = \mathbf{A} \cdot \cdot \mathbf{1}$$

## REMARK

$\mathbf{A} : \mathbf{B} \neq \mathbf{A} \cdot \cdot \mathbf{B}$   
Unless one of the two  
tensors is symmetric.

# Example

□ Prove that:

$$\mathbf{A} : \mathbf{B} = \text{Tr}(\mathbf{A}^T \cdot \mathbf{B})$$

$$\mathbf{A} \cdot \mathbf{B} = \text{Tr}(\mathbf{A} \cdot \mathbf{B})$$

# Example - Solution

$$c = \text{Tr}(\mathbf{A}^T \cdot \mathbf{B}) = [\mathbf{A}^T \cdot \mathbf{B}]_{kk} = [\mathbf{A}^T]_{ki} [\mathbf{B}]_{ik} = \underbrace{A_{ik} B_{ik}}_{k \rightarrow j} = A_{ij} B_{ij}$$

$$c = \text{Tr}(\mathbf{A} \cdot \mathbf{B}) = [\mathbf{A} \cdot \mathbf{B}]_{kk} = [\mathbf{A}]_{ki} [\mathbf{B}]_{ik} = \underbrace{A_{ki} B_{ik}}_{\substack{i \rightarrow j \\ k \rightarrow i}} = A_{ij} B_{ji}$$

# 2<sup>nd</sup> Order Tensor Operations

## □ **Determinant** yields a scalar

$$\det \mathbf{A} = \det [\mathbf{A}] = \det \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix} = \epsilon_{ijk} A_{1i} A_{2j} A_{3k} = \frac{1}{6} \epsilon_{ijk} \epsilon_{pqr} A_{pi} A_{qj} A_{rk}$$

### ■ Some **properties**:

$$\det(\mathbf{A} \cdot \mathbf{B}) = \det \mathbf{A} \cdot \det \mathbf{B}$$

$$\det \mathbf{A}^T = \det \mathbf{A}$$

$$\det(\alpha \mathbf{A}) = \alpha^3 \det \mathbf{A}$$

### REMARK

The tensor  $\mathbf{A}$  is **SINGULAR** if and only if  $\det \mathbf{A} = 0$ .

$\mathbf{A}$  is **NONSINGULAR** if  $\det \mathbf{A} \neq 0$ .

## □ **Inverse**

There exists a **unique inverse**  $\mathbf{A}^{-1}$  of  $\mathbf{A}$  when  $\mathbf{A}$  is nonsingular, which satisfies the reciprocal relation:

$$\begin{cases} \mathbf{A} \cdot \mathbf{A}^{-1} = \mathbf{1} = \mathbf{A}^{-1} \cdot \mathbf{A} \\ A_{ik} A_{kj}^{-1} = A_{ik}^{-1} A_{kj} = \delta_{ij} \quad i, j, k \in \{1, 2, 3\} \end{cases}$$

# 2<sup>nd</sup> Order Tensor Operations

- If **A** and **B** are **invertible**, the following **properties** apply:

$$(\mathbf{A} \cdot \mathbf{B})^{-1} = \mathbf{B}^{-1} \cdot \mathbf{A}^{-1}$$

$$(\mathbf{A}^{-1})^{-1} = \mathbf{A}$$

$$(\alpha \mathbf{A})^{-1} = \frac{1}{\alpha} \mathbf{A}^{-1}$$

$$(\mathbf{A}^{-1})^T = (\mathbf{A}^T)^{-1} = \mathbf{A}^{-T}$$

$$\mathbf{A}^{-2} = \mathbf{A}^{-1} \cdot \mathbf{A}^{-1}$$

$$\det(\mathbf{A}^{-1}) = (\det \mathbf{A})^{-1} = 1/\det \mathbf{A}$$

# Example

□ Prove that  $\det \mathbf{A} = \epsilon_{ijk} A_{1i} A_{2j} A_{3k}$



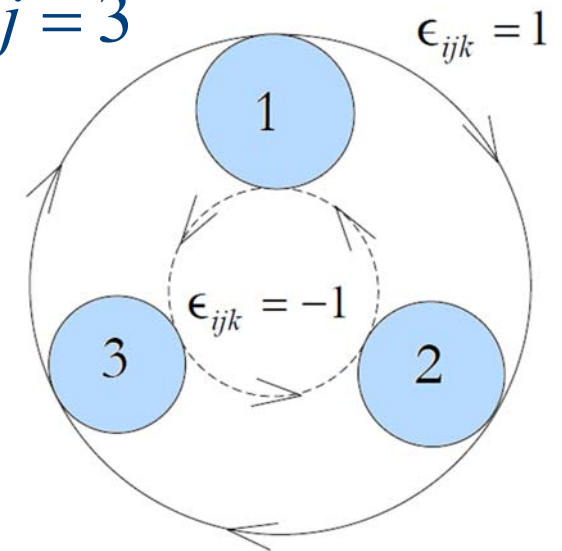
# Example - Solution

$$\det \mathbf{A} = \det \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}$$

$$= A_{11}A_{22}A_{33} + A_{21}A_{32}A_{13} + A_{31}A_{12}A_{23} - A_{13}A_{22}A_{31} - A_{23}A_{32}A_{11} - A_{33}A_{12}A_{21}$$

# Example - Solution

$$\begin{aligned}
 \epsilon_{ijk} A_{1i} A_{2j} A_{3k} &= \cancel{\epsilon_{111} A_{11} A_{21} A_{31}} + \cancel{\epsilon_{112} A_{11} A_{21} A_{32}} + \cancel{\epsilon_{113} A_{11} A_{21} A_{33}} + & j=1 \\
 &+ \cancel{\epsilon_{121} A_{11} A_{22} A_{31}} + \cancel{\epsilon_{122} A_{11} A_{22} A_{32}} + \epsilon_{123} A_{11} A_{22} A_{33} + & j=2 \\
 &+ \cancel{\epsilon_{131} A_{11} A_{23} A_{31}} + \epsilon_{132} A_{11} A_{23} A_{32} + \cancel{\epsilon_{133} A_{11} A_{23} A_{33}} + & j=3 \\
 &+ \cancel{\epsilon_{211} A_{12} A_{21} A_{31}} + \cancel{\epsilon_{212} A_{12} A_{21} A_{32}} + \epsilon_{213} A_{12} A_{21} A_{33} + \\
 &+ \cancel{\epsilon_{221} A_{12} A_{22} A_{31}} + \cancel{\epsilon_{222} A_{12} A_{22} A_{32}} + \cancel{\epsilon_{223} A_{12} A_{22} A_{33}} + \\
 &+ \epsilon_{231} A_{12} A_{23} A_{31} + \cancel{\epsilon_{232} A_{12} A_{23} A_{32}} + \cancel{\epsilon_{233} A_{12} A_{23} A_{33}} + \\
 &+ \cancel{\epsilon_{311} A_{13} A_{21} A_{31}} + \epsilon_{312} A_{13} A_{21} A_{32} + \cancel{\epsilon_{313} A_{13} A_{21} A_{33}} + \\
 &+ \epsilon_{321} A_{13} A_{22} A_{31} + \cancel{\epsilon_{322} A_{13} A_{22} A_{32}} + \cancel{\epsilon_{323} A_{13} A_{22} A_{33}} + \\
 &+ \cancel{\epsilon_{331} A_{13} A_{23} A_{31}} + \cancel{\epsilon_{332} A_{13} A_{23} A_{32}} + \cancel{\epsilon_{333} A_{13} A_{23} A_{33}} = \\
 &= A_{11} A_{22} A_{33} + A_{12} A_{23} A_{31} + A_{13} A_{21} A_{32} - A_{13} A_{22} A_{31} - A_{12} A_{21} A_{33} - A_{11} A_{23} A_{32}
 \end{aligned}$$



# Summary - Tensor Operations

- ▣ **Dot product** – contraction of one index:

$$c = \mathbf{u} \cdot \mathbf{v} = u_i v_i$$

$$[\mathbf{C}]_{ij} = [\mathbf{A} \cdot \mathbf{B}]_{ij} = A_{ik} B_{kj} = C_{ij}$$

$$[\mathbf{c}]_j = [\mathbf{u} \cdot \mathbf{A}]_j = u_i A_{ij} = c_j$$

$$[\mathbf{d}]_i = [\mathbf{A} \cdot \mathbf{u}]_i = A_{ij} u_j = d_i$$

$$[\mathcal{C}]_{ijk} = [\mathbf{A} \cdot \mathcal{B}]_{ijk} = A_{im} B_{mjk} = \mathcal{C}_{ijk}$$

$$[\mathcal{D}]_{ijk} = [\mathcal{B} \cdot \mathbf{A}]_{ijk} = B_{ijm} A_{mk} = \mathcal{D}_{ijk}$$

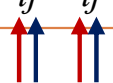
$$[\mathbf{D}]_{ij} = [\mathbf{B} \cdot \mathbf{A}]_{ij} = B_{ik} A_{kj} = D_{ij}$$

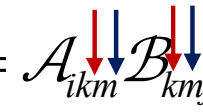
$$[\mathbb{C}]_{ijkl} = [\mathcal{A} \cdot \mathcal{B}]_{ijkl} = \mathcal{A}_{ijm} \mathcal{B}_{mkl} = \mathbb{C}_{ijkl}$$


$$[\mathbb{D}]_{ijkl} = [\mathcal{B} \cdot \mathcal{A}]_{ijkl} = \mathcal{B}_{ijm} \mathcal{A}_{mkl} = \mathbb{D}_{ijkl}$$

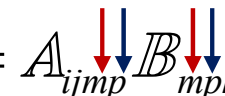
# Summary - Tensor Operations


- ▣ **Double dot product** – contraction of two indices:

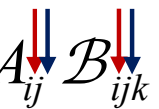
$$c = \mathbf{A} : \mathbf{B} = A_{ij} B_{ij}$$


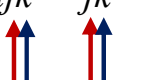
$$[\mathbf{C}]_{ij} = [\mathbf{A} : \mathbf{B}]_{ij} = A_{ikm} B_{kmj} = C_{ij}$$



$$[\mathbf{D}]_{ij} = [\mathbf{B} : \mathbf{A}]_{ij} = B_{ikm} A_{kmj} = D_{ij}$$


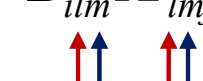
$$[\mathbb{C}]_{ijkl} = [\mathbf{A} : \mathbf{B}]_{ijkl} = A_{ijmp} B_{mpkl} = \mathbb{C}_{ijkl}$$


$$[\mathbb{D}]_{ijkl} = [\mathbf{B} : \mathbf{A}]_{ijkl} = B_{ijmp} A_{mpij} = \mathbb{D}_{ijkl}$$


$$[\mathbf{c}]_k = [\mathbf{A} : \mathbf{B}]_k = A_{ij} B_{ijk} = c_k$$


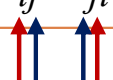
$$[\mathbf{d}]_i = [\mathbf{B} : \mathbf{A}]_i = B_{ijk} A_{jk} = d_i$$



$$[\mathbb{C}]_{ijk} = [\mathbf{A} : \mathbf{B}]_{ijk} = A_{ijlm} B_{lmk} = \mathbb{C}_{ijk}$$


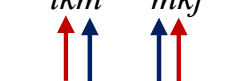
$$[\mathbb{D}]_{ijk} = [\mathbf{B} : \mathbf{A}]_{ijk} = B_{ilm} A_{lmjk} = \mathbb{D}_{ijk}$$


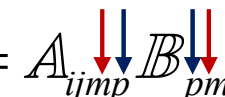
# Summary - Tensor Operations


- ▣ **Transposed double dot product** – contraction of two indexes:


$$c = \mathbf{A} \cdot \cdot \mathbf{B} = A_{ij} B_{ji}$$



$$[\mathbf{C}]_{ij} = [\mathcal{A} \cdot \cdot \mathcal{B}]_{ij} = \mathcal{A}_{ikm} \mathcal{B}_{mkj} = C_{ij}$$


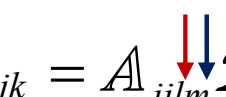
$$[\mathbf{D}]_{ij} = [\mathcal{B} \cdot \cdot \mathcal{A}]_{ij} = \mathcal{B}_{ikm} \mathcal{A}_{mkj} = D_{ij}$$


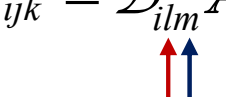
$$[\mathbb{C}]_{ijkl} = [\mathbf{A} \cdot \cdot \mathbf{B}]_{ijkl} = A_{ijmp} B_{pmkl} = \mathbb{C}_{ijkl}$$


$$[\mathbb{D}]_{ijkl} = [\mathcal{B} \cdot \cdot \mathcal{A}]_{ijkl} = \mathcal{B}_{ijmp} \mathcal{A}_{pmij} = \mathbb{D}_{ijkl}$$


$$[\mathbf{c}]_k = [\mathbf{A} \cdot \cdot \mathcal{B}]_k = A_{ij} \mathcal{B}_{jik} = c_k$$


$$[\mathbf{d}]_i = [\mathcal{B} \cdot \cdot \mathbf{A}]_i = \mathcal{B}_{ijk} A_{kj} = d_i$$


$$[\mathcal{C}]_{ijk} = [\mathbf{A} \cdot \cdot \mathcal{B}]_{ijk} = A_{ijlm} \mathcal{B}_{mlk} = \mathcal{C}_{ijk}$$


$$[\mathcal{D}]_{ijk} = [\mathcal{B} \cdot \cdot \mathbf{A}]_{ijk} = \mathcal{B}_{ilm} A_{mljk} = \mathcal{D}_{ijk}$$


# Summary - Tensor Operations

▣ **Open product** – expansion of indexes:

$$\begin{aligned} [\mathbf{A}]_{ij} &= [\mathbf{u} \otimes \mathbf{v}]_{ij} = u_i v_j = A_{ij} \\ [\mathbf{C}]_{ijkl} &= [\mathbf{A} \otimes \mathbf{B}]_{ijkl} = A_{ij} B_{kl} = \mathcal{C}_{ijkl} \end{aligned}$$

$$\begin{aligned} [\mathbf{B}]_{ij} &= [\mathbf{v} \otimes \mathbf{u}]_{ij} = v_i u_j = B_{ij} \\ [\mathbb{D}]_{ijkl} &= [\mathbf{B} \otimes \mathbf{A}]_{ijkl} = B_{ij} A_{kl} = \mathbb{D}_{ijkl} \end{aligned}$$

$$[\mathcal{C}]_{ijk} = [\mathbf{u} \otimes \mathbf{A}]_{ijk} = u_i A_{jk} = \mathcal{C}_{ijk}$$

$$[\mathcal{D}]_{ijk} = [\mathbf{A} \otimes \mathbf{u}]_{ijk} = A_{ij} u_k = \mathcal{D}_{ijk}$$

$$[\mathfrak{C}]_{ijklm} = [\mathbf{A} \otimes \mathcal{B}]_{ijklm} = A_{ij} \mathcal{B}_{klm} = \mathfrak{C}_{ijklm}$$

$$[\mathfrak{D}]_{ijklm} = [\mathcal{B} \otimes \mathbf{A}]_{ijklm} = \mathcal{B}_{ijk} A_{lm} = \mathfrak{D}_{ijklm}$$

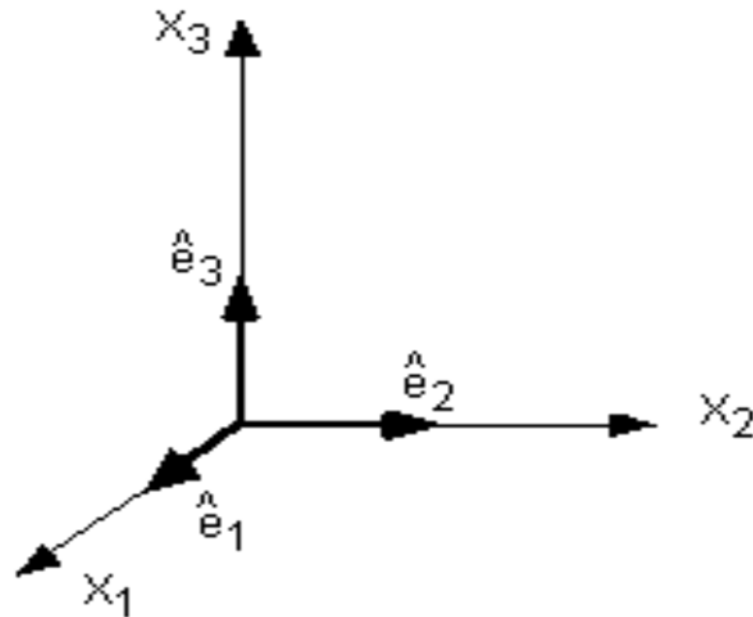


# Differential Operators

Tensor Algebra

# Differential Operators

- A **differential operator** is a mapping that transforms a field  $\mathbf{v}(\mathbf{x})$ ,  $\mathbf{A}(\mathbf{x})\dots$  into another field by means of partial derivatives.
  - The mapping is typically understood to be linear.
  - Examples:
    - Nabla operator
    - Gradient
    - Divergence
    - Rotation
    - ...





# Nabla Operator

- ▣ The **Nabla operator**  $\nabla$  is a differential operator “symbolically” defined as:

$$\nabla \stackrel{\text{symbolic}}{=} \frac{\partial}{\partial \mathbf{x}} \stackrel{\text{symb.}}{=} \frac{\partial}{\partial x_i} \hat{\mathbf{e}}_i$$

- ▣ In Cartesian coordinates, it can be used as a (symbolic) vector on its own:

$$[\nabla] \stackrel{\text{symb.}}{=} \begin{bmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \frac{\partial}{\partial x_3} \end{bmatrix}$$

# Gradient

- ▣ The **gradient** (or **open product of Nabla**) is a differential operator defined as:

- Gradient of a scalar field  $\Phi(\mathbf{x})$ :

- Yields a vector

$$\left\{ \begin{array}{l} [\nabla \Phi]_i = [\nabla \otimes \Phi]_i = [\nabla]_i \Phi \stackrel{\text{symp.}}{=} \frac{\partial}{\partial x_i} \Phi = \frac{\partial \Phi}{\partial x_i} \quad i \in \{1, 2, 3\} \\ \nabla \Phi = [\nabla \Phi]_i \hat{\mathbf{e}}_i = \frac{\partial \Phi}{\partial x_i} \hat{\mathbf{e}}_i \end{array} \right.$$

$$\nabla \Phi = \frac{\partial \Phi}{\partial x_i} \hat{\mathbf{e}}_i$$

- Gradient of a vector field  $\mathbf{v}(\mathbf{x})$ :

- Yields a 2<sup>nd</sup> order tensor

$$\left\{ \begin{array}{l} [\nabla \otimes \mathbf{v}]_{ij} = [\nabla]_i [\mathbf{v}]_j \stackrel{\text{symp.}}{=} \frac{\partial}{\partial x_i} v_j = \frac{\partial v_j}{\partial x_i} \quad i, j \in \{1, 2, 3\} \\ \nabla \mathbf{v} = \nabla \otimes \mathbf{v} = [\nabla \otimes \mathbf{v}]_{ij} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j = \frac{\partial v_j}{\partial x_i} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \end{array} \right.$$

$$\nabla \mathbf{v} = \frac{\partial v_j}{\partial x_i} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j$$

# Gradient

- Gradient of a 2<sup>nd</sup> order tensor field  $\mathbf{A}(\mathbf{x})$ :

- Yields a 3<sup>rd</sup> order tensor

$$\left\{ \begin{array}{l} [\nabla \mathbf{A}]_{ijk} = [\nabla \otimes \mathbf{A}]_{ijk} = [\nabla]_i [\mathbf{A}]_{jk} \stackrel{\text{symp.}}{=} \frac{\partial}{\partial x_i} A_{jk} = \frac{\partial A_{jk}}{\partial x_i} \quad i, j, k \in \{1, 2, 3\} \\ \nabla \mathbf{A} = \nabla \otimes \mathbf{A} = [\nabla \otimes \mathbf{A}]_{ijk} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k = \frac{\partial A_{jk}}{\partial x_i} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k \end{array} \right.$$

$$\nabla \mathbf{A} = \frac{\partial A_{jk}}{\partial x_i} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_j \otimes \hat{\mathbf{e}}_k$$

# Divergence

- ▣ The **divergence** (or **dot product of Nabla**) is a differential operator defined as :

- Divergence of a vector field  $\mathbf{v}(\mathbf{x})$ :

- Yields a scalar

$$\nabla \cdot \mathbf{v} = [\nabla]_i [\mathbf{v}]_i \stackrel{\text{symp.}}{=} \frac{\partial}{\partial x_i} v_i = \frac{\partial v_i}{\partial x_i}$$

$$\nabla \cdot \mathbf{v} = \frac{\partial v_i}{\partial x_i}$$

- Divergence of a 2<sup>nd</sup> order tensor  $\mathbf{A}(\mathbf{x})$ :

- Yields a vector

$$\begin{cases} [\nabla \cdot \mathbf{A}]_j = [\nabla]_i [\mathbf{A}]_{ij} \stackrel{\text{symp.}}{=} \frac{\partial}{\partial x_i} A_{ij} = \frac{\partial A_{ij}}{\partial x_i} & j \in \{1, 2, 3\} \\ \nabla \cdot \mathbf{A} = [\nabla \cdot \mathbf{A}]_j \hat{\mathbf{e}}_j = \frac{\partial A_{ij}}{\partial x_i} \hat{\mathbf{e}}_j \end{cases}$$

$$\nabla \cdot \mathbf{A} = \frac{\partial A_{ij}}{\partial x_i} \hat{\mathbf{e}}_j$$

# Divergence

- ▣ The divergence can only be performed on tensors of order 1 or higher.
- ▣ If  $\nabla \cdot \mathbf{v} = 0$ , the vector field  $\mathbf{v}(\mathbf{x})$  is said to be **solenoid** (or **divergence-free**).

# Rotation

- The **rotation** or **curl** (or **vector product of Nabla**) is a differential operator defined as:

- Rotation of a vector field  $\mathbf{v}(\mathbf{x})$ :

- Yields a vector

$$\left\{ \begin{array}{l} [\nabla \times \mathbf{v}]_i \stackrel{\text{symp.}}{=} \epsilon_{ijk} [\nabla]_j [\mathbf{v}]_k \stackrel{\text{symp.}}{=} \epsilon_{ijk} \frac{\partial}{\partial x_j} v_k = \epsilon_{ijk} \frac{\partial v_k}{\partial x_j} \quad i \in \{1, 2, 3\} \\ \nabla \times \mathbf{v} = [\nabla \times \mathbf{v}]_i \hat{\mathbf{e}}_i = \epsilon_{ijk} \frac{\partial v_k}{\partial x_j} \hat{\mathbf{e}}_i \end{array} \right.$$

$$\nabla \times \mathbf{v} = \epsilon_{ijk} \frac{\partial v_k}{\partial x_j} \hat{\mathbf{e}}_i$$

- Rotation of a 2<sup>nd</sup> order tensor  $\mathbf{A}(\mathbf{x})$ :

- Yields a 2<sup>nd</sup> order tensor



$$\left\{ \begin{array}{l} [\nabla \times \mathbf{A}]_{il} \stackrel{\text{symp.}}{=} \epsilon_{ijk} \frac{\partial}{\partial x_j} A_{kl} = \epsilon_{ijk} \frac{\partial A_{kl}}{\partial x_j} \quad i, j, k \in \{1, 2, 3\} \\ \nabla \times \mathbf{A} = [\nabla \times \mathbf{A}]_{il} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_l = \epsilon_{ijk} \frac{\partial A_{kl}}{\partial x_j} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_l \end{array} \right.$$

$$\nabla \times \mathbf{A} = \epsilon_{ijk} \frac{\partial A_{kl}}{\partial x_j} \hat{\mathbf{e}}_i \otimes \hat{\mathbf{e}}_l$$

# Rotation

- ▣ The rotation can only be performed on tensors of order 1 or higher.
- ▣ If  $\nabla \times \mathbf{v} = 0$ , the vector field  $\mathbf{v}(\mathbf{x})$  is said to be **irrotational** (or **curl-free**).

# Differential Operators - Summary

	scalar field $\Phi(\mathbf{x})$	vector field $\mathbf{v}(\mathbf{x})$	2 <sup>nd</sup> order tensor $\mathbf{A}(\mathbf{x})$
<b>GRADIENT</b>	$[\nabla \otimes \Phi]_i = [\nabla \Phi]_i = \frac{\partial \Phi}{\partial x_i}$	$[\nabla \otimes \mathbf{v}]_{ij} = [\nabla \mathbf{v}]_{ij} = \frac{\partial v_j}{\partial x_i}$	$[\nabla \otimes \mathbf{A}]_{ijk} = [\nabla \mathbf{A}]_{ijk} = \frac{\partial A_{jk}}{\partial x_i}$
<b>DIVERGENCE</b>		$\nabla \cdot \mathbf{v} = \frac{\partial v_i}{\partial x_i}$	$[\nabla \cdot \mathbf{A}]_j = \frac{\partial A_{ij}}{\partial x_i}$
<b>ROTATION</b>		$[\nabla \times \mathbf{v}]_i = \epsilon_{ijk} \frac{\partial v_k}{\partial x_j}$	$[\nabla \times \mathbf{A}]_{il} = \epsilon_{ijk} \frac{\partial A_{kl}}{\partial x_j}$



# Example

- Given the vector  $\mathbf{v} = \mathbf{v}(\mathbf{x}) = x_1 x_2 x_3 \hat{\mathbf{e}}_1 + x_1 x_2 \hat{\mathbf{e}}_2 + x_1 \hat{\mathbf{e}}_3$  determine  $\nabla \cdot \mathbf{v}$ ,  $\nabla \times \mathbf{v}$ ,  $\nabla \mathbf{v}$ .

# Example - Solution

$$\mathbf{v} = \mathbf{v}(\mathbf{x}) = x_1 x_2 x_3 \hat{\mathbf{e}}_1 + x_1 x_2 \hat{\mathbf{e}}_2 + x_1 \hat{\mathbf{e}}_3$$



$$[\mathbf{v}] = \begin{bmatrix} x_1 x_2 x_3 \\ x_1 x_2 \\ x_1 \end{bmatrix}$$

▣ Divergence:

$$\nabla \cdot \mathbf{v} = \frac{\partial v_i}{\partial x_i}$$

$$\nabla \cdot \mathbf{v} = \frac{\partial v_i}{\partial x_i} = \frac{\partial v_1}{\partial x_1} + \frac{\partial v_2}{\partial x_2} + \frac{\partial v_3}{\partial x_3} = x_2 x_3 + x_1$$

# Example - Solution

□ Divergence:

$$\nabla \cdot \mathbf{v} = \frac{\partial v_i}{\partial x_i}$$

$$[\mathbf{v}] = \begin{bmatrix} x_1 x_2 x_3 \\ x_1 x_2 \\ x_1 \end{bmatrix}$$

■ In matrix notation:

$$\begin{aligned} \nabla \cdot \mathbf{v} &= [\nabla]^T [\mathbf{v}] = \underbrace{\begin{bmatrix} \frac{\partial}{\partial x_1} & \frac{\partial}{\partial x_2} & \frac{\partial}{\partial x_3} \end{bmatrix}}_{1 \times 3} \underbrace{\begin{bmatrix} x_1 x_2 x_3 \\ x_1 x_2 \\ x_1 \end{bmatrix}}_{3 \times 1} = \\ &= \underbrace{\frac{\partial}{\partial x_1} x_1 x_2 x_3 + \frac{\partial}{\partial x_2} x_1 x_2 + \frac{\partial}{\partial x_3} x_1}_{1 \times 1} = \frac{\partial (x_1 x_2 x_3)}{\partial x_1} + \frac{\partial (x_1 x_2)}{\partial x_2} + \frac{\partial x_1}{\partial x_3} = x_2 x_3 + x_1 \end{aligned}$$

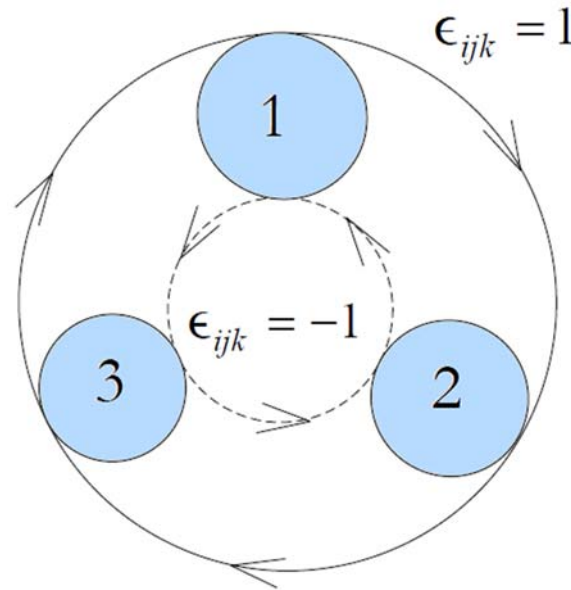
# Example - Solution

□ Rotation:

$$[\nabla \times \mathbf{v}]_i = \epsilon_{ijk} \frac{\partial v_k}{\partial x_j}$$

■ In index notation:

$$\begin{aligned} [\nabla \times \mathbf{v}]_i &= \epsilon_{ijk} \frac{\partial v_k}{\partial x_j} = \\ &= \epsilon_{i12} \frac{\partial v_2}{\partial x_1} + \epsilon_{i13} \frac{\partial v_3}{\partial x_1} + \epsilon_{i21} \frac{\partial v_1}{\partial x_2} + \epsilon_{i23} \frac{\partial v_3}{\partial x_2} + \epsilon_{i31} \frac{\partial v_1}{\partial x_3} + \epsilon_{i32} \frac{\partial v_2}{\partial x_3} = \end{aligned}$$



$$[\mathbf{v}] = \begin{bmatrix} x_1 x_2 x_3 \\ x_1 x_2 \\ x_1 \end{bmatrix}$$

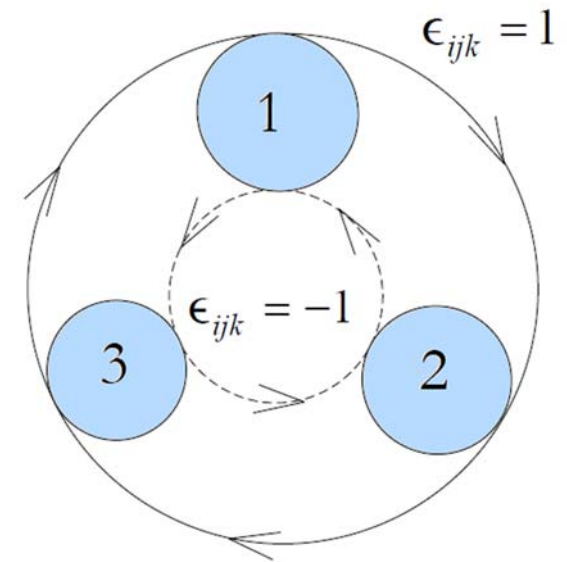
# Example - Solution

□ Rotation: 
$$[\nabla \times \mathbf{v}]_i = \epsilon_{i12} \frac{\partial v_2}{\partial x_1} + \epsilon_{i13} \frac{\partial v_3}{\partial x_1} + \epsilon_{i21} \frac{\partial v_1}{\partial x_2} + \epsilon_{i23} \frac{\partial v_3}{\partial x_2} + \epsilon_{i31} \frac{\partial v_1}{\partial x_3} + \epsilon_{i32} \frac{\partial v_2}{\partial x_3}$$

$$[\mathbf{v}] = \begin{bmatrix} x_1 x_2 x_3 \\ x_1 x_2 \\ x_1 \end{bmatrix}$$

■ In matrix notation

$$[\nabla \times \mathbf{v}] = \begin{bmatrix} \underbrace{\epsilon_{123}}_{=1} \frac{\partial v_3}{\partial x_2} + \underbrace{\epsilon_{132}}_{=-1} \frac{\partial v_2}{\partial x_3} \\ \underbrace{\epsilon_{213}}_{=-1} \frac{\partial v_3}{\partial x_1} + \underbrace{\epsilon_{231}}_{=1} \frac{\partial v_1}{\partial x_3} \\ \underbrace{\epsilon_{312}}_{=1} \frac{\partial v_2}{\partial x_1} + \underbrace{\epsilon_{321}}_{=-1} \frac{\partial v_1}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 0 \\ x_1 x_2 - 1 \\ x_2 - x_1 x_3 \end{bmatrix}$$



■ In compact notation:  $\nabla \times \mathbf{v} = (x_1 x_2 - 1) \hat{\mathbf{e}}_2 + (x_2 - x_1 x_3) \hat{\mathbf{e}}_3$

# Example - Solution

## □ Rotation:

### ■ Calculated directly in matrix notation:

$$\begin{aligned} \nabla \times \mathbf{v} &= \begin{bmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \frac{\partial}{\partial x_3} \end{bmatrix} \times \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} = \det \begin{bmatrix} \hat{\mathbf{e}}_1 & \hat{\mathbf{e}}_2 & \hat{\mathbf{e}}_3 \\ \frac{\partial}{\partial x_1} & \frac{\partial}{\partial x_2} & \frac{\partial}{\partial x_3} \\ v_1 & v_2 & v_3 \end{bmatrix} = \\ &= \left( \frac{\partial v_3}{\partial x_2} - \frac{\partial v_2}{\partial x_3} \right) \hat{\mathbf{e}}_1 + \left( \frac{\partial v_1}{\partial x_3} - \frac{\partial v_3}{\partial x_1} \right) \hat{\mathbf{e}}_2 + \left( \frac{\partial v_2}{\partial x_1} - \frac{\partial v_1}{\partial x_2} \right) \hat{\mathbf{e}}_3 = \\ &= (x_1 x_2 - 1) \hat{\mathbf{e}}_2 + (x_2 - x_1 x_3) \hat{\mathbf{e}}_3 \end{aligned}$$

$$v_1 = x_1 x_2 x_3$$

$$v_2 = x_1 x_2$$

$$v_3 = x_1$$

# Example - Solution

## ▣ Gradient:

$$[\nabla \otimes \mathbf{v}]_{ij} = [\nabla \mathbf{v}]_{ij} = \frac{\partial v_j}{\partial x_i}$$

$$v_1 = x_1 x_2 x_3$$

$$v_2 = x_1 x_2$$

$$v_3 = x_1$$

## ■ In matrix notation

$$[\nabla \mathbf{v}] \stackrel{\text{symp}}{=} [\nabla \otimes \mathbf{v}] \stackrel{\text{symp}}{=} [\nabla][\mathbf{v}]^T \stackrel{\text{symp}}{=} \underbrace{\begin{bmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \frac{\partial}{\partial x_3} \end{bmatrix}}_{3 \times 1} \underbrace{\begin{bmatrix} x_1 x_2 x_3 & x_1 x_2 & x_1 \end{bmatrix}}_{1 \times 3} = \underbrace{\begin{bmatrix} x_2 x_3 & x_2 & 1 \\ x_1 x_3 & x_1 & 0 \\ x_1 x_2 & 0 & 0 \end{bmatrix}}_{3 \times 3}$$

## ■ In compact notation:

$$\nabla \mathbf{v} = x_2 x_3 \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_1 + x_2 \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_2 + \hat{\mathbf{e}}_1 \otimes \hat{\mathbf{e}}_3 + x_1 x_3 \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_1 + x_1 \hat{\mathbf{e}}_2 \otimes \hat{\mathbf{e}}_2 + x_1 x_2 \hat{\mathbf{e}}_3 \otimes \hat{\mathbf{e}}_1$$



# Integral Theorems

Tensor Algebra



# Divergence or Gauss Theorem

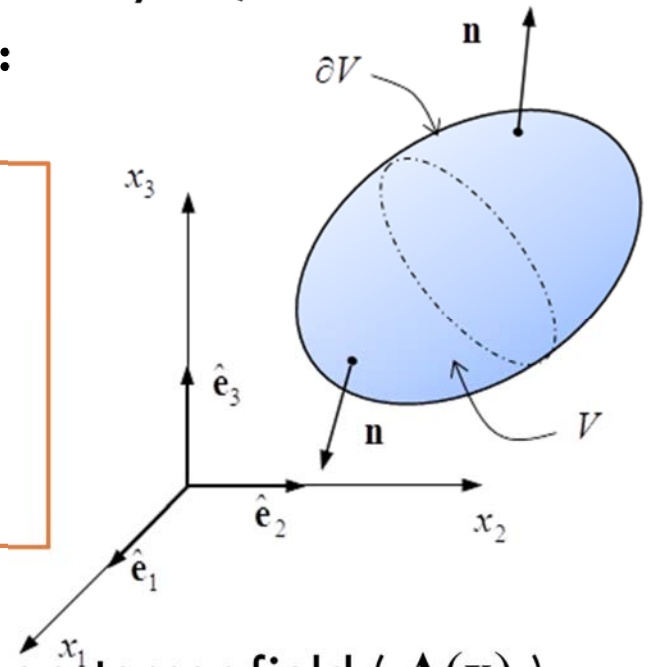
- Given a field  $\mathcal{A}$  in a volume  $V$  with closed boundary surface  $\partial V$  and unit **outward** normal to the boundary  $\mathbf{n}$ , the **Divergence (or Gauss) Theorem** states:

$$\int_V \nabla \cdot \mathcal{A} \, dV = \int_{\partial V} \mathbf{n} \cdot \mathcal{A} \, dS$$

$$\int_V \mathcal{A} \cdot \nabla \, dV = \int_{\partial V} \mathcal{A} \cdot \mathbf{n} \, dS$$

Where:

- $\mathcal{A}$  represents either a vector field (  $\mathbf{v}(\mathbf{x})$  ) or a tensor field (  $\mathbf{A}(\mathbf{x})$  ).



# Generalized Divergence Theorem

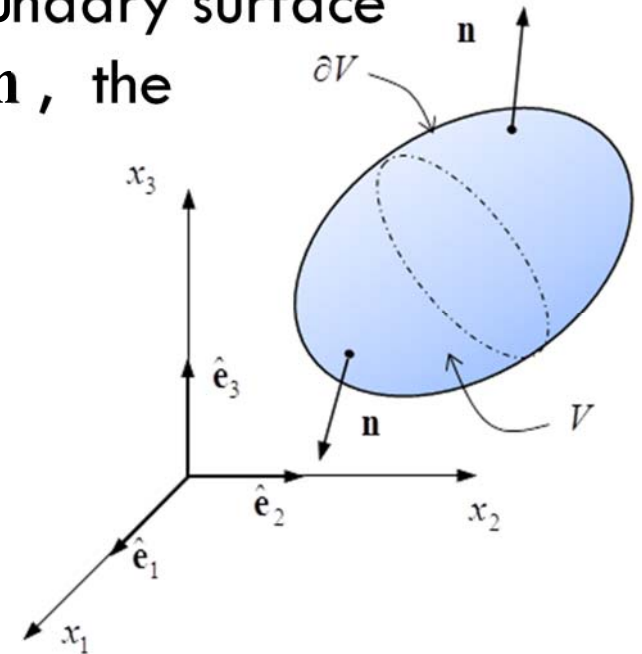
- Given a field  $\mathcal{A}$  in a volume  $V$  with closed boundary surface  $\partial V$  and unit outward normal to the boundary  $\mathbf{n}$ , the **Generalized Divergence Theorem** states:

$$\int_V \nabla * \mathcal{A} dV = \int_{\partial V} \mathbf{n} * \mathcal{A} dS$$

$$\int_V \mathcal{A} * \nabla dV = \int_{\partial V} \mathcal{A} * \mathbf{n} dS$$

Where:

- $*$  represents either the dot product (  $\cdot$  ), the cross product (  $\times$  ) or the tensor product (  $\otimes$  ).
- $\mathcal{A}$  represents either a scalar field (  $\phi(\mathbf{x})$  ), a vector field (  $\mathbf{v}(\mathbf{x})$  ) or a tensor field (  $\mathbf{A}(\mathbf{x})$  ).

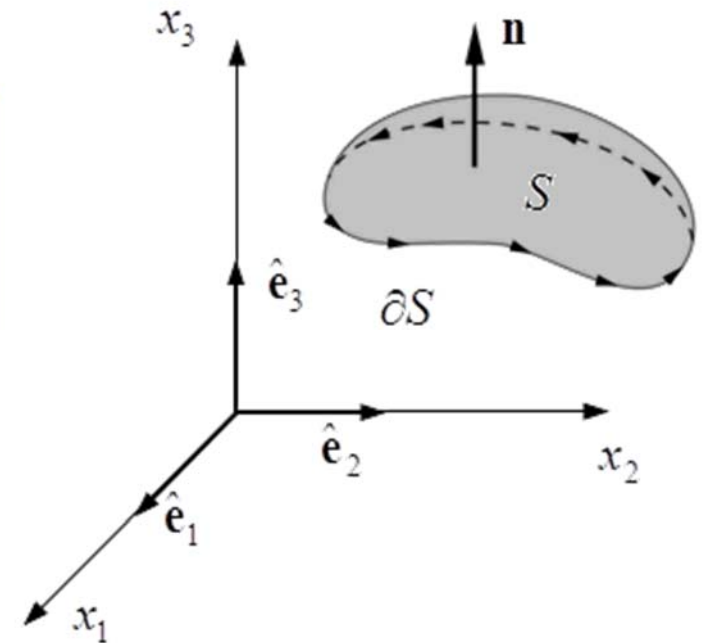


# Curl or Stokes Theorem

- Given a vector field  $\mathbf{u}$  in a surface  $S$  with closed boundary surface  $\partial S$  and unit outward normal to the boundary  $\mathbf{n}$ , the **Curl (or Stokes) Theorem** states:

$$\int_S (\nabla \times \mathbf{u}) \cdot \mathbf{n} dS = \int_{\partial S} \mathbf{u} \cdot d\mathbf{r}$$

where the curve of the line integral must have positive orientation, such that  $d\mathbf{r}$  points counter-clockwise when the unit normal points to the viewer, following the right-hand rule.



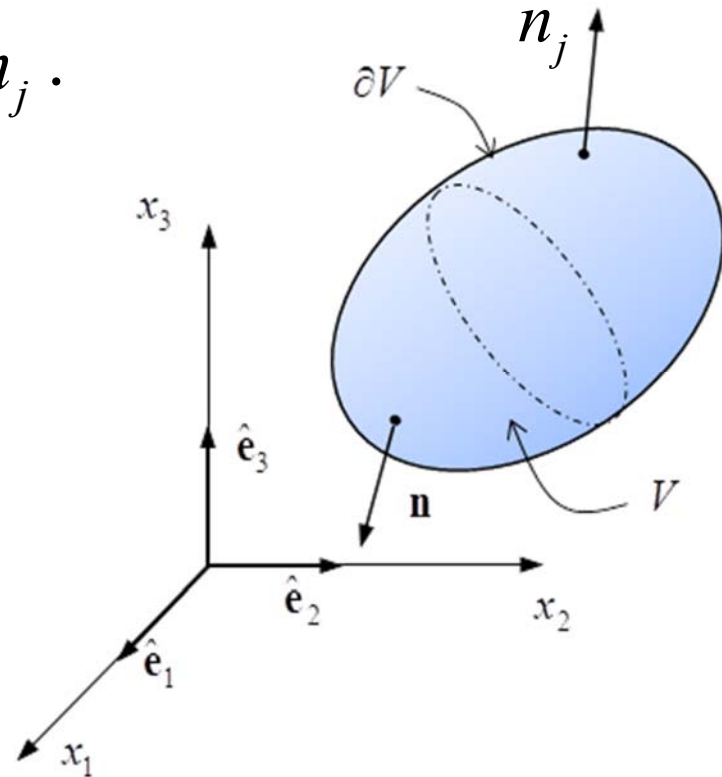
# Example

- Use the Generalized Divergence Theorem to show that

$$\int_S x_i n_j dS = V \delta_{ij}$$

where  $x_i$  is the position vector of  $n_j$ .

$$\int_{\partial V} \mathcal{A} * \mathbf{n} dS = \int_V \mathcal{A} * \nabla dV$$



# Example - Solution

$$\int_S x_i n_j dS \quad \Rightarrow \quad \int_S \mathbf{x} \otimes \mathbf{n} dS$$

$$\int_S x_i n_j dS = V \delta_{ij}$$

- ▣ Applying the Generalized Divergence Theorem:

$$\int_{\partial V} \mathbf{x} \otimes \mathbf{n} dS = \int_V \mathbf{x} \otimes \nabla dV$$

- ▣ Applying the definition of gradient of a vector:

$$[\nabla \mathbf{x}]_{ij} = \frac{\partial x_j}{\partial x_i} \quad \Rightarrow \quad [\mathbf{x} \nabla]_{ij} = \frac{\partial x_i}{\partial x_j}$$

# Example - Solution

- The Generalized Divergence Theorem in index notation:

$$\int_S x_i n_j dS = V \delta_{ij}$$

$$\int_S x_i n_j dS = \int_V \frac{\partial x_i}{\partial x_j} dV$$

- Then,

$$\int_S x_i n_j dS = \int_V \frac{\partial x_i}{\partial x_j} dV = \int_V \delta_{ij} dV = \delta_{ij} V$$



# References

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# References



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