

Introduction to Tensors

Algebra of second order tensors

A second order tensor \mathbf{A} is a linear transformation mapping of a vector to another vector, i.e.

$$\mathbf{u} = \mathbf{A}\mathbf{v}.$$

As \mathbf{A} is a linear transformation, it implies

$$\mathbf{A}(\alpha\mathbf{u} + \beta\mathbf{v}) = \alpha\mathbf{A}\mathbf{u} + \beta\mathbf{A}\mathbf{v}.$$

The *tensor product* or *the dyad* of two vectors is a second order tensor defined as,

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

Note that the dot product is between the two immediate adjacent vectors which are not connected by \otimes symbol.

Sometimes dyad is simply written as \mathbf{uv} .

It also follows, $(\alpha\mathbf{u} + \beta\mathbf{v}) \otimes \mathbf{w} = \alpha\mathbf{uw} + \beta\mathbf{vw}$

Another relation,

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

Notice the vectors for which dot product is taken.

A second order tensor can also be represented as a dyadic or tensor product of cartesian basis vectors \mathbf{e}_i ($i \in [1, 2, 3]$) as,

$$\mathbf{A} = A_{ij} \mathbf{e}_i \otimes \mathbf{e}_j \text{ or } \mathbf{A} = A_{ij} \mathbf{e}_i \mathbf{e}_j$$

where $\mathbf{e}_i \mathbf{e}_j$ may be thought as a ‘base tensor’ in terms of which tensor \mathbf{A} may be expanded in Cartesian frame. It is analogous to a vector (first order tensor) being expanded in terms of ‘base vectors’ \mathbf{e}_i .

The components of second order tensor \mathbf{A} in a particular coordinate system can be represented as a 3×3 matrix as:

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

The components of a *unit* or *identity* tensor \mathbf{I} is represented as:

$$\mathbf{I} = \delta_{ij} \mathbf{e}_i \mathbf{e}_j = \mathbf{e}_i \mathbf{e}_i$$

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

Examples

Example 5:

For a given second order tensor \mathbf{A} find the $(m,n)^{\text{th}}$ component of the tensor.

The $(m,n)^{\text{th}}$ component of tensor \mathbf{A} can be extracted by post-multiplying with \mathbf{e}_n , and then pre-multiplying with \mathbf{e}_m as,

$$\begin{aligned}\mathbf{e}_m \cdot \mathbf{A} \mathbf{e}_n &= \mathbf{e}_m \cdot (A_{ij} \mathbf{e}_i \otimes \mathbf{e}_j) \mathbf{e}_n \\ &\Rightarrow \mathbf{e}_m \cdot (A_{ij} \mathbf{e}_i \otimes \mathbf{e}_j) \mathbf{e}_n \\ &\Rightarrow \mathbf{e}_m \cdot (A_{ij} \mathbf{e}_j \cdot \mathbf{e}_n) \mathbf{e}_i \\ &\Rightarrow \mathbf{e}_m \cdot (A_{ij} \delta_{jn}) \mathbf{e}_i \\ &\Rightarrow A_{ij} \delta_{jn} \mathbf{e}_m \cdot \mathbf{e}_i \\ &\Rightarrow A_{in} \mathbf{e}_m \cdot \mathbf{e}_i \\ &\Rightarrow A_{in} \delta_{mi} \\ &\Rightarrow A_{mn}\end{aligned}$$

Examples

Example 6:

Show that $\mathbf{v} = \mathbf{A}\mathbf{u}$ in the tensorial form can be written as

$$v_i = A_{ij}u_j.$$

We start by writing \mathbf{A} as $A_{ij}\mathbf{e}_i\mathbf{e}_j$ and \mathbf{u} as $u_k\mathbf{e}_k$, then

$$\begin{aligned}\mathbf{A}\mathbf{u} &= (A_{ij}\mathbf{e}_i\mathbf{e}_j)(u_k\mathbf{e}_k) \\ &\Rightarrow A_{ij}u_k(\mathbf{e}_j \cdot \mathbf{e}_k)\mathbf{e}_i \\ &\Rightarrow A_{ij}u_k\delta_{jk}\mathbf{e}_i \\ &\Rightarrow A_{ij}u_j\mathbf{e}_i = v_i\mathbf{e}_i\end{aligned}$$

Thus, $v_i = A_{ij}u_j$

Transpose of a tensor

The *transpose* of a tensor \mathbf{A} is denoted by \mathbf{A}^T and is defined as,

$$\mathbf{A}^T = A_{ji}\mathbf{e}_i\mathbf{e}_j \text{ or } (\mathbf{A}^T)_{ij} = A_{ji}.$$

Definition of tranpose is governed by the following identity. For any two vector \mathbf{u} and \mathbf{v} ,

$$\mathbf{v} \cdot \mathbf{A}^T \mathbf{u} = \mathbf{u} \cdot \mathbf{A} \mathbf{v} = \mathbf{A} \mathbf{v} \cdot \mathbf{u}.$$

Proof: $\mathbf{v} \cdot \mathbf{A}^T \mathbf{u} = (v_m \mathbf{e}_m) \cdot (A_{ji} \mathbf{e}_i \mathbf{e}_j) u_k \mathbf{e}_k$ $u_j A_{ji} v_i = \mathbf{u} \cdot \mathbf{A} \mathbf{v}$
 $\Rightarrow (v_m \mathbf{e}_m) \cdot A_{ji} u_k \delta_{jk} \mathbf{e}_i$ or
 $\Rightarrow A_{ji} u_k v_m \delta_{jk} \delta_{mi}$ $A_{ji} v_i u_j = \mathbf{A} \mathbf{v} \cdot \mathbf{u}$
 $\Rightarrow A_{ji} u_j v_i$

From the definition following identities immidialtely follow,

$$(\mathbf{A}^T)^T = \mathbf{A}, \quad (\mathbf{A} \mathbf{B})^T = \mathbf{B}^T \mathbf{A}^T, \quad (\mathbf{u} \otimes \mathbf{v})^T = \mathbf{v} \otimes \mathbf{u}$$

Contraction

- *Contraction* is an operation in which we identify two indices and sum over them. Contraction is characterized as a dot.
- *Double contraction* or scalar product of two tensors \mathbf{A} and \mathbf{B} is characterized as two dots and yield a scalar,

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

Proof: $\mathbf{A} : \mathbf{B} = (A_{ij} \mathbf{e}_i \mathbf{e}_j) : (B_{kl} \mathbf{e}_k \mathbf{e}_l)$
 $\Rightarrow A_{ij} B_{kl} \delta_{ik} \delta_{jl} = A_{ij} B_{ij}$

(Notice the order in which dot product of basis vectors are taken)

- *Double contraction* of any tensors \mathbf{A} with identity tensor yields the *trace* of tensor \mathbf{A} .

$$\mathbf{A} : \mathbf{I} = A_{ij} \delta_{ij} = A_{ii} = \text{tr} \mathbf{A} = A_{11} + A_{22} + A_{33}$$

Examples

Example 7: Show that $\mathbf{A} : (\mathbf{B} \cdot \mathbf{C}) = (\mathbf{B}^T \cdot \mathbf{A}) : \mathbf{C} = (\mathbf{A} \cdot \mathbf{C}^T) : \mathbf{B}$.

Let's start from LHS:

$$\begin{aligned}\mathbf{A} : (\mathbf{B} \cdot \mathbf{C}) &= A_{mn} \mathbf{e}_m \mathbf{e}_n : (B_{ij} \mathbf{e}_i \mathbf{e}_j \cdot C_{kl} \mathbf{e}_k \mathbf{e}_l) \\ &\Rightarrow A_{mn} \mathbf{e}_m \mathbf{e}_n : B_{ij} C_{kl} \mathbf{e}_i \mathbf{e}_l \delta_{jk} \\ &\Rightarrow A_{mn} \mathbf{e}_m \mathbf{e}_n : B_{ik} C_{kl} \mathbf{e}_i \mathbf{e}_l \\ &\Rightarrow A_{mn} B_{ik} C_{kl} \delta_{mi} \delta_{nl} \\ &\Rightarrow A_{mn} B_{mk} C_{kn}\end{aligned}$$

Above can also be written as following,

$$\begin{aligned}A_{mn} B_{mk} C_{kn} &= B_{mk} A_{mn} C_{kn} = (\mathbf{B}^T \cdot \mathbf{A}) : \mathbf{C} \\ A_{mn} B_{mk} C_{kn} &= A_{mn} C_{kn} B_{mk} = (\mathbf{A} \cdot \mathbf{C}^T) : \mathbf{B}\end{aligned}$$

Examples

Example 8: Show that $(\mathbf{A} \otimes \mathbf{B}) : \mathbf{C} = \mathbf{A}(\mathbf{B} : \mathbf{C})$.

Let's start from LHS:

$$\begin{aligned}(\mathbf{A} \otimes \mathbf{B}) : \mathbf{C} &= (A_{mn} \mathbf{e}_m \mathbf{e}_n \otimes B_{ij} \mathbf{e}_i \mathbf{e}_j) : C_{kl} \mathbf{e}_k \mathbf{e}_l \\&\Rightarrow A_{mn} \mathbf{e}_m \mathbf{e}_n B_{ij} C_{kl} \delta_{ik} \delta_{jl} \\&\Rightarrow A_{mn} \mathbf{e}_m \mathbf{e}_n B_{kl} C_{kl} = \mathbf{A}(\mathbf{B} : \mathbf{C})\end{aligned}$$

Determinant and Inverse of a tensor

The *determinant* of a second order tensor is a scalar and is given as

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

with properties,

$$\det(\mathbf{AB}) = \det \mathbf{A} \det \mathbf{B} \text{ and } \det(\mathbf{A}^T) = \det(\mathbf{A}).$$

A tensor \mathbf{A} is said to be singular *if and only if* $\det(\mathbf{A})=0$

For a non-singular tensor \mathbf{A} , there exists a unique inverse tensor \mathbf{A}^{-1} such that,

$$\mathbf{AA}^{-1} = \mathbf{A}^{-1}\mathbf{A} = \mathbf{I}$$

Inverse of a tensor

Invertible tensors have the following important properties:

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

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

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

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

An *orthogonal tensor* is a special tensor whose inverse is same as its transpose, i.e.

$$\mathbf{Q}^T = \mathbf{Q}^{-1},$$

which follows, $\mathbf{Q}\mathbf{Q}^T = \mathbf{Q}^T\mathbf{Q} = \mathbf{I}$.

$$\text{Also, } \det(\mathbf{Q}^T\mathbf{Q}) = (\det \mathbf{Q})^2 = 1.$$

If $\det \mathbf{Q} = +1$ tensor is called *proper* orthogonal tensor, and if $\det \mathbf{Q} = -1$ it is called *improper* orthogonal tensor.

Symmetric and skew tensor

A second order tensor is symmetric if $\mathbf{S} = \mathbf{S}^T$ or $S_{ij} = S_{ji}$.

A tensor is called skew or antisymmetric if $\mathbf{W} = -\mathbf{W}^T$ or $W_{ij} = -W_{ji}$.

Any tensor \mathbf{A} can be decomposed into a symmetric and skew tensor as,

$$\mathbf{A} = \mathbf{S} + \mathbf{W}$$

where,

$$\mathbf{S} = \frac{\mathbf{A} + \mathbf{A}^T}{2}, \text{ and } \mathbf{W} = \frac{\mathbf{A} - \mathbf{A}^T}{2},$$

which have following forms,

$$[\mathbf{S}] = \begin{pmatrix} S_{11} & S_{12} & S_{13} \\ S_{12} & S_{22} & S_{23} \\ S_{13} & S_{23} & S_{33} \end{pmatrix} \quad \text{and} \quad [\mathbf{W}] = \begin{pmatrix} 0 & W_{12} & W_{13} \\ -W_{12} & 0 & W_{23} \\ -W_{13} & -W_{23} & 0 \end{pmatrix}_{24}$$

Examples

Example 9: If \mathbf{S} is a symmetric, and \mathbf{W} is an antisymmetric tensor, then show that

(i) $\mathbf{S} : \mathbf{W} = 0$,

(ii) $\mathbf{S} : \mathbf{B} = \mathbf{S} : \text{symm}(\mathbf{B})$, and

(iii) $\mathbf{W} : \mathbf{B} = \mathbf{W} : \text{asymm}(\mathbf{B})$

where \mathbf{B} is a second order tensor; $\text{symm}(\mathbf{B})$ and $\text{asymm}(\mathbf{B})$ are *symmetric* and *antisymmetric* part of \mathbf{B} , respectively.

Consider the fact that, for any second order tensors \mathbf{S} and \mathbf{W} , we can write,

$$\mathbf{S} : \mathbf{W} = S_{ij}W_{ij} = S_{ji}W_{ji}.$$

$$\mathbf{S} : \mathbf{W} = 1/2 (S_{ij}W_{ij} + S_{ij}W_{ij}) = 1/2 (S_{ij}W_{ij} + S_{ji}W_{ji})$$

As \mathbf{S} is symmetric and \mathbf{W} is skew, $S_{ij} = S_{ji}$, and $W_{ij} = -W_{ji}$.

Hence, we can write, $\mathbf{S} : \mathbf{W} = 1/2 (S_{ij}W_{ij} - S_{ij}W_{ij}) = 0$.

Now, tensor \mathbf{B} can be splitted in to symmetric and antisymmetric part, hence, $\mathbf{B} = \text{symm}(\mathbf{B}) + \text{asymm}(\mathbf{B})$

$$\mathbf{S} : \mathbf{B} = \mathbf{S} : (\text{symm}(\mathbf{B}) + \text{asymm}(\mathbf{B})) = \mathbf{S} : \text{symm}(\mathbf{B}),$$

as $\mathbf{S} : \text{asymm}(\mathbf{B}) = 0$.

Similarly,

$$\mathbf{W} : \mathbf{B} = \mathbf{W} : (\text{symm}(\mathbf{B}) + \text{asymm}(\mathbf{B})) = \mathbf{W} : \text{asymm}(\mathbf{B}),$$

as $\mathbf{W} : \text{symm}(\mathbf{B}) = 0$.

Spherical and deviatoric tensor

Any tensors \mathbf{A} can be split into a spherical and a deviatoric part as,

$$\mathbf{A} = \alpha \mathbf{I} + \text{dev} \mathbf{A} \text{ or } A_{ij} = \alpha \delta_{ij} + \text{dev} A_{ij},$$

where, scalar α is given as $\alpha = \frac{1}{3} \text{tr} \mathbf{A} = \frac{1}{3} A_{ii}$.

Deviatoric part is calculated as,

$$\text{dev} A_{ij} = A_{ij} - \alpha \delta_{ij} = A_{ij} - \frac{1}{3} A_{kk} \delta_{ij}.$$

Deviatoric tensor has an important property that,

$$\text{tr}(\text{dev} \mathbf{A}) = (\text{dev} \mathbf{A})_{mm} = A_{mm} - \frac{1}{3} A_{kk} \delta_{mm} = A_{mm} - A_{kk} = 0$$

Thus trace of any deviatoric tensor is always *zero*.

Transformation laws of tensors

- Through a tensor is invariant with respect to coordinate system, its components change if the coordinate system changes.
- Consider two coordinate system denoted by x_i and \bar{x}_i with base vectors e_i and \bar{e}_i , respectively.
- As shown in the figure the components of a vector \mathbf{v} are $v_i = \mathbf{e}_i \cdot \mathbf{v}$ in the first system, and $\bar{v}_i = \bar{\mathbf{e}}_i \cdot \mathbf{v}$ in the second system.
- Similarly, the components of a tensor \mathbf{A} is $A_{ij} = \mathbf{e}_i \cdot \mathbf{A} \mathbf{e}_j$ in the first system and \bar{A} is $\bar{A}_{ij} = \bar{\mathbf{e}}_i \cdot \mathbf{A} \bar{\mathbf{e}}_j$ in the second system. We will derive the transformation laws for a tensor, i.e. the relationship between the components A_{ij} and \bar{A}_{ij} .

