

Preliminaries

A generic vector \underline{v} can be expressed as:

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

where $\underline{e}_1, \underline{e}_2$ and \underline{e}_3 are the unit vectors of a Cartesian coordinate system, and v_1, v_2 and v_3 are the components of \underline{v} along the directions 1, 2 and 3.

The following notation can be adopted:

$$\underline{v} = v_i \underline{e}_i \quad i = 1, 2, 3$$

where the summation is implied with respect to the repeated indexes (in this case just i). In other words:

$$\begin{aligned} \underline{v} &= \sum_{i=1}^3 v_i \underline{e}_i \\ &= v_i \underline{e}_i \end{aligned}$$

Einstein's convention

Similarly, a second-order tensor can be represented as:

$$\underline{\underline{A}} = A_{ij} \underline{e}_i \otimes \underline{e}_j$$

where summation is now implied with respect to i and j , or:

$$\begin{aligned} \underline{\underline{A}} &= \sum_{i=1}^3 \sum_{j=1}^3 A_{ij} \underline{e}_i \otimes \underline{e}_j \\ &= A_{ij} \underline{e}_i \otimes \underline{e}_j \end{aligned}$$

The term A_{ij} identifies the components of the tensor \underline{A} for a given set of unit vectors \underline{e}_i .

The tensor itself does not depend upon the choice of the reference system, but its components A_{ij} depend on the reference system.

$$\begin{aligned}\underline{A} &= A_{ij} \underline{e}_i \otimes \underline{e}_j \\ &= \hat{A}_{ij} \hat{\underline{e}}_i \otimes \hat{\underline{e}}_j\end{aligned} \quad \leadsto \quad A_{ij} \neq \hat{A}_{ij}$$

where \underline{e}_i and $\hat{\underline{e}}_i$ are different sets of unit vectors.

Often it is useful to collect the components of vectors and second-order tensors into one- and two-dimensional arrays:

$$\underline{v} = v_i \underline{e}_i \quad \leadsto \quad \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$$

$$\underline{A} = A_{ij} \underline{e}_i \otimes \underline{e}_j \quad \leadsto \quad \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}$$

This is an intuitive way for "storing" the components of a tensor, making it possible to perform tensor operations by means of the techniques of matrix algebra.

However it is worth highlighting that tensors are not arrays!

Scalar product

The scalar product between two vectors a and b is:

$$c = \underline{a} \cdot \underline{b}$$

$$= (a_1 \underline{e}_1 + a_2 \underline{e}_2 + a_3 \underline{e}_3) \cdot (b_1 \underline{e}_1 + b_2 \underline{e}_2 + b_3 \underline{e}_3)$$

$$= a_1 b_1 \underline{e}_1 \cdot \underline{e}_1 + a_1 b_2 \cancel{\underline{e}_1 \cdot \underline{e}_2} + a_1 b_3 \cancel{\underline{e}_1 \cdot \underline{e}_3} + \\ a_2 b_1 \cancel{\underline{e}_2 \cdot \underline{e}_1} + a_2 b_2 \underline{e}_2 \cdot \underline{e}_2 + a_2 b_3 \cancel{\underline{e}_2 \cdot \underline{e}_3} + \\ a_3 b_1 \cancel{\underline{e}_3 \cdot \underline{e}_1} + a_3 b_2 \cancel{\underline{e}_3 \cdot \underline{e}_2} + a_3 b_3 \underline{e}_3 \cdot \underline{e}_3$$

$$= a_1 b_1 \underline{e}_1 \cdot \underline{e}_1 + a_2 b_2 \underline{e}_2 \cdot \underline{e}_2 + a_3 b_3 \underline{e}_3 \cdot \underline{e}_3$$

$$= a_1 b_1 + a_2 b_2 + a_3 b_3$$

Using a compact notation:

$$c = \underline{a} \cdot \underline{b}$$

$$= a_i \underline{e}_i \cdot b_j \underline{e}_j$$

$$= a_i b_j \underline{e}_i \cdot \underline{e}_j$$

$$= a_i b_j \delta_{ij}$$

$$= a_i b_i = a_j b_j$$

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

Note: the indexes used for a and b must be different(!) to avoid any misunderstanding on the repeated indexes.

Consider now the scalar product between a second order tensor and a vector:

$$\underline{\underline{A}} \cdot \underline{b} = \underline{\underline{A}} \underline{b} \quad (\text{the dot can be explicitly reported or not})$$

$$= A_{ij} \underline{e}_i \otimes \underline{e}_j \cdot b_k \underline{e}_k$$

$$= A_{ij} b_k \underline{e}_i \otimes \underbrace{\underline{e}_j \cdot \underline{e}_k}_{\delta_{jk}}$$

$$= A_{ij} b_k \underline{e}_i \delta_{jk}$$

$$= A_{ij} b_k \delta_{jk} \underline{e}_i$$

$$= A_{ij} b_j \underline{e}_i = A_{ik} b_k \underline{e}_i$$

The result is then a vector \underline{c} , whose component c_i is equal to $A_{ik} b_k$.

Expanding the previous expression:

$$\underline{\underline{A}} \cdot \underline{b} = A_{ik} b_k \underline{e}_i$$

$$= A_{1k} b_k \underline{e}_1 + A_{2k} b_k \underline{e}_2 + A_{3k} b_k \underline{e}_3$$

$$= (A_{11} b_1 + A_{12} b_2 + A_{13} b_3) \underline{e}_1 +$$

$$(A_{21} b_1 + A_{22} b_2 + A_{23} b_3) \underline{e}_2 +$$

$$(A_{31} b_1 + A_{32} b_2 + A_{33} b_3) \underline{e}_3$$

Double scalar product (contraction)

When dealing with second- (or higher-) order tensors it is possible to introduce a "double" scalar product, which is commonly denoted with a double dot.

$$\underline{a} \cdot \underline{b} = c$$

(scalar product)

A scalar quantity is obtained from two vectors.

$$\underline{A} : \underline{B} = c$$

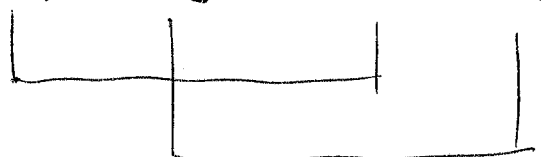
(contraction)

A scalar quantity is obtained from two second-order tensors.

$$\underline{A} : \underline{B} = A_{ij} \underline{e}_i \otimes \underline{e}_j : B_{rs} \underline{e}_r \otimes \underline{e}_s$$

$$= A_{ij} B_{rs} \underline{e}_i \otimes \underline{e}_j : \underline{e}_r \otimes \underline{e}_s$$

It is important to establish a convention to define how the double scalar product is taken (this choice is not unique, and different conventions are found in the literature)

$$= A_{ij} B_{rs} \underline{e}_i \otimes \underline{e}_j : \underline{e}_r \otimes \underline{e}_s$$


The convention adopted here considers the scalar products according to the graphical representation above, meaning that:

$$= A_{ij} B_{rs} (\underline{e}_i \cdot \underline{e}_r) (\underline{e}_j \cdot \underline{e}_s)$$

$$= A_{ij} B_{rs} \delta_{ir} \delta_{js}$$

$$= A_{ij} B_{ij} = A_{rs} B_{rs}$$

The expanded form is then:

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

$$= A_{11} B_{11} + A_{21} B_{21} + A_{31} B_{31}$$

$$+ A_{12} B_{12} + A_{22} B_{22} + A_{32} B_{32} +$$

$$A_{13} B_{13} + A_{23} B_{23} + A_{33} B_{33}$$

One example of application of the contraction operation is the evaluation of the strain energy density:

$$u = \frac{1}{2} \underline{\underline{\sigma}} : \underline{\underline{\epsilon}}$$

where $\underline{\underline{\sigma}}$ and $\underline{\underline{\epsilon}}$ are proper measures of stresses and strains. (which are second order tensors)

One can note that if $\underline{\underline{\sigma}}$ and $\underline{\underline{\epsilon}}$ are symmetric then the contraction reads:

$$\begin{aligned} \underline{\underline{\sigma}} : \underline{\underline{\epsilon}} &= \sigma_{11} \epsilon_{11} + \sigma_{12} \epsilon_{12} + \sigma_{13} \epsilon_{13} + \\ &\quad \sigma_{12} \epsilon_{12} + \sigma_{22} \epsilon_{22} + \sigma_{23} \epsilon_{23} + \\ &\quad \sigma_{13} \epsilon_{13} + \sigma_{23} \epsilon_{23} + \sigma_{33} \epsilon_{33} \\ &= \sigma_{11} \epsilon_{11} + \sigma_{22} \epsilon_{22} + \sigma_{33} \epsilon_{33} + \\ &\quad \sigma_{12} 2 \epsilon_{12} + \sigma_{13} 2 \epsilon_{13} + \sigma_{23} 2 \epsilon_{23} \end{aligned}$$

For this reason, stress and strain components can be collected into one-dimensional arrays as:

$$\underline{\underline{\sigma}} \rightsquigarrow \begin{pmatrix} \sigma_{11} \\ \sigma_{22} \\ \sigma_{33} \\ \sigma_{23} \\ \sigma_{13} \\ \sigma_{12} \end{pmatrix} \quad \underline{\underline{\epsilon}} \rightsquigarrow \begin{pmatrix} \epsilon_{11} \\ \epsilon_{22} \\ \epsilon_{33} \\ 2 \epsilon_{23} \\ 2 \epsilon_{13} \\ 2 \epsilon_{12} \end{pmatrix}$$

and the operation of contraction is then represented as:

$$\underline{\underline{\sigma}} : \underline{\underline{\epsilon}} \rightarrow \{ \epsilon_{11} \quad \epsilon_{22} \quad \epsilon_{33} \quad 2\epsilon_{23} \quad 2\epsilon_{13} \quad 2\epsilon_{12} \} \begin{Bmatrix} \sigma_{11} \\ \sigma_{22} \\ \sigma_{33} \\ \sigma_{23} \\ \sigma_{13} \\ \sigma_{12} \end{Bmatrix}$$

$$= \underline{\underline{\epsilon}}^T \underline{\underline{\sigma}} = \underline{\underline{\sigma}}^T \underline{\underline{\epsilon}}$$

This notation, generally denoted as Voigt-notation, is based on the representation of second-order tensor components by means of one-dimensional arrays. While it is useful for performing several algebraic manipulations, it is important to bear in mind that $\underline{\underline{\sigma}}$ and $\underline{\underline{\epsilon}}$ are not vectors

For instance, the rotation of $\underline{\underline{\sigma}}$ and $\underline{\underline{\epsilon}}$ follows the rules of the rotation of a second-order tensor, not those of a vector!

• Gradient and divergence

The gradient and divergence operators are defined as:

$\nabla(\cdot) = \text{grad}(\cdot) = \frac{\partial}{\partial x_i}(\cdot) \underline{e}_i$	<u>gradient</u>
$\nabla \cdot (\cdot) = \text{div}(\cdot) = \frac{\partial}{\partial x_i}(\cdot) \cdot \underline{e}_i$	<u>divergence</u>

According to the above definitions, the gradient of a vector \underline{a} is:

$$\begin{aligned}\text{grad } \underline{a} &= \frac{\partial \underline{a}}{\partial x_i} \underline{e}_i \\&= \frac{\partial a_j}{\partial x_i} \underline{e}_j \underline{e}_i \\&= \frac{\partial a_j}{\partial x_i} \underline{e}_j \otimes \underline{e}_i \\&= a_{j/i} \underline{e}_j \otimes \underline{e}_i = a_{i/j} \underline{e}_i \otimes \underline{e}_j\end{aligned}$$

Similarly, the divergence is:

$$\begin{aligned}\text{div } \underline{a} &= \frac{\partial \underline{a}}{\partial x_i} \cdot \underline{e}_i \\&= \frac{\partial a_j}{\partial x_i} \underline{e}_j \cdot \underline{e}_i \\&= \frac{\partial a_j}{\partial x_i} \delta_{ij} = a_{i/i}\end{aligned}$$

The divergence of a second-order tensor is:

$$\operatorname{div} \underline{\underline{A}} = \frac{\partial A_{rs}}{\partial x_i} \underline{e}_r \otimes \underline{e}_s \cdot \underline{e}_i$$

$$= A_{rs/i} \delta_{si} \underline{e}_r$$

$$= A_{ri/i} \underline{e}_r = A_{ir/r} \underline{e}_i$$