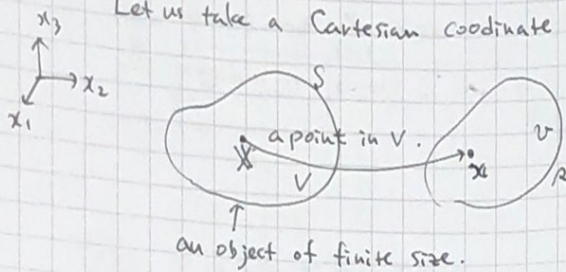


(1) What is continuum?

(2) displacement

Let us take a Cartesian coordinate  $x_1, x_2$  &  $x_3$  in  $\mathbb{R}^3$ .



The object deforms as  $V \rightarrow v$ .

(1)  $u_i = x_i - X_i$  is called displacement.

In the following, the "map"  $X \mapsto x$  is "one-to-one" i.e. we can define the "inverse"  $x \mapsto X$  and the map is sufficiently smooth (i.e. we can define

(3) Deformation gradient tensor.

Let us consider a "short" vector  $dX = Y - X$  embedded in  $V$ , and assume that

$dX$  becomes as  $dx$  after the deformation.

$$\frac{d^n}{dx_1 dx_2 \dots dx_n} x(X) \text{ for arbitrary } n$$

We can use the map  $x_i(X)$  to evaluate the 1st order approx. of  $dx_i$  as

$$\begin{aligned} dx_i &= x_i(Y) - x_i(X) \\ &= x_i(X + dX) - x_i(X) \\ &\approx x_i(X) + \frac{\partial x_i(X)}{\partial X_j} dX_j - x_i(X) \quad (2) \\ &= \frac{\partial x_i}{\partial X_j} dX_j \leftarrow \text{Einstein's summation conv. is used.} \end{aligned}$$

Let us henceforth denote  $\frac{\partial x_i}{\partial X_j}$  as  $F_{ij}$ , and call  $F_{ij}$  as "deformation gradient tensor".

$$\text{Q1} \quad F_{ij} = \frac{\partial}{\partial X_j} (x_i) = \frac{\partial}{\partial X_j} (X_i + u_i) = \delta_{ij} + \frac{\partial u_i}{\partial X_j}$$

$$\begin{aligned} \det(F) &= \det(dX \cdot (dY \times dZ)) \\ &= dx \cdot (dy \times dz) \end{aligned}$$

ij comp. of

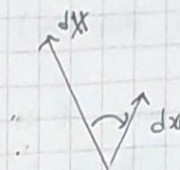
Q2. If  $F_{ij}$  is orthogonal matrix,

$$\begin{aligned} dx_i dx_i &= F_{ij} dX_j F_{ik} dX_k \\ &= dX_j F_{ji}^T F_{ik} dX_k = dX_j dX_j \\ &= \delta_{jk} \end{aligned}$$



$$dx = F dx$$

$F$  describes the rotation of  $dx$  as well as the deformation.



→ We want to decompose  $F$  into two parts.  
deformation & rotation.

Theorem. For any  $F$  of positive definite matrix, there exist

$\left\{ \begin{array}{l} R : \text{orthogonal matrix} \\ U \text{ \& } V : \text{positive definite \& symmetric matrix} \end{array} \right.$   
such that  $F = RU = VR \dots (3)$   
Also, such  $R, U$  &  $V$  are unique.

$R$  : Rotation

First, we prove the following lemma:

lemma For any positive definite & symmetric matrix  $A$ , there exist positive definite & symmetric matrix  $B$  such that  $A = B^2$ , and such a  $B$  is unique.

( $\because$ )  $A$  can be diagonalized with an orthogonal matrix  $P$  as

$$P^T A P = \Lambda = \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{pmatrix} \dots (4)$$

where  $\lambda_i > 0$  because  $A$  is positive definite.

Let us define  $B$  as  $B = P \begin{pmatrix} \sqrt{\lambda_1} & 0 & 0 \\ 0 & \sqrt{\lambda_2} & 0 \\ 0 & 0 & \sqrt{\lambda_3} \end{pmatrix} P^T$ , then

it is obvious that  $B^2 = P \begin{pmatrix} & & \\ & & \\ & & \end{pmatrix} P^T P \begin{pmatrix} & & \\ & & \\ & & \end{pmatrix} P^T = P \Lambda P^T = A \dots (5)$

we have proved the existence of  $B$ .

We then need to show the uniqueness. To this end let us define

another P.D & S matrix  $\tilde{B}$  such that  $A = \tilde{B}^2$ , &  $\tilde{B} \neq B$ .

Let us take an eigenpair  $(\tilde{\lambda}, \tilde{p})$  of  $\tilde{B}$ , to see

$$\tilde{B} \tilde{p} = \tilde{\lambda} \tilde{p} \Leftrightarrow \tilde{B}^2 \tilde{p} = \tilde{B} \tilde{\lambda} \tilde{p} = \tilde{\lambda} \tilde{p}$$

$$\Leftrightarrow A \tilde{p} = \tilde{\lambda} \tilde{p}.$$

Thus,  $A$  &  $\tilde{B}$  share the eigenvector, which means that  $P$  in (4)

diagonalize  $\tilde{B}$  as well as  $P^T \tilde{B} P = \begin{pmatrix} \tilde{\lambda}_1 & & \\ & \tilde{\lambda}_2 & \\ & & \tilde{\lambda}_3 \end{pmatrix}$ .  $\tilde{\lambda}_i^2 = \lambda_i \Rightarrow \tilde{\lambda}_i = \lambda_i$ .

$$\Leftrightarrow P^T \tilde{B}^2 P = \begin{pmatrix} \tilde{\lambda}_1 & & \\ & \tilde{\lambda}_2 & \\ & & \tilde{\lambda}_3 \end{pmatrix} \dots (b) \quad \therefore \tilde{B} = B.$$

(Proof of the main theorem).

First note that, when  $F$  is positive definite,  $F^T F$  is also positive definite. This can be seen as

$$\det F^T F = \det F^T \det F = (\det F)^2 > 0. \\ (\because F \text{ is positive definite})$$

$F^T F$  is also symmetric because

$$(F^T F)_{ij} = F_{ik}^T F_{kj} = F_{ki} F_{kj} = F_{ki} F_{jk}^T = F_{jk}^T F_{ki} = (F^T F)_{ji}$$

holds. According to the lemma, for  $F^T F$ , there is a unique positive definite & symmetric matrix  $U$  such that

$$F^T F = U^2 \quad \text{Exist because } U \text{ is positive def.}$$

Then, we can define  $R := F U^{-1} \Leftrightarrow F = R U$ .  $\star$

$$R^T R = U^{-1} F^T F U^{-1} = I \quad (\because U^2 = F^T F)$$

The factorisation  $\star$  is unique because of the uniqueness of  $U$ .

( $F = VR$  can be proved in the similar manner.  
starting from  $FF^T$ , which is also p.d.)

Note:  $C = F^T F$  is called right Cauchy Green tensor

$B = FF^T$  is left



Let us now compute the length of  $d\mathbf{x}$  &  $d\mathbf{x}$ .

14

$$|d\mathbf{x}|^2 = dx_i dx_i$$

$$= F_{ij} dx_j F_{ik} dx_k$$

$$= dx_j F_{ji}^T F_{ik} dx_k$$

$$= dx_j C_{jk} dx_k$$

$$|d\mathbf{x}|^2 - |d\mathbf{X}|^2 = dx_j C_{jk} dx_k - dx_j dx_j$$

$$= dx_j C_{jk} dx_k - dx_j \delta_{jk} dx_k$$

$$= dx_j (C_{jk} - \delta_{jk}) dx_k$$

$$= d\mathbf{X} \cdot (\mathbf{C} - \mathbf{I}) d\mathbf{X}$$

Note:  $\mathbf{C} - \mathbf{I} = \mathbf{F}^T \mathbf{F} - \mathbf{I} = \mathbf{U}^2 - \mathbf{I}$   $\leftarrow$  does not involve  $\mathbf{R}$  (rotation)

$\mathbf{E} = \frac{1}{2}(\mathbf{C} - \mathbf{I})$  Green's strain tensor.

On the other hand,

$$|d\mathbf{x}|^2 - |d\mathbf{X}|^2 = dx_j dx_j - F_{ij}^{-1} dx_j F_{ik}^{-1} dx_k$$

$$= dx_j (\delta_{jk} - F_{ji}^{-t} F_{ik}^{-1}) dx_k$$

$$= dx_j (\delta_{jk} - (F F^t)^{-1})_{jk} dx_k$$

$$= d\mathbf{x} \cdot (\mathbf{I} - \mathbf{B}^{-1}) d\mathbf{x}$$

$\mathbf{I} - \mathbf{B}^{-1} = \mathbf{I} - \mathbf{V}^2$  does not involve  $\mathbf{R}$  either

$\mathbf{e} = \frac{1}{2}(\mathbf{I} - \mathbf{B}^{-1})$  : Almansi's strain tensor

infinitesimal deformation theory.  $F_{ij} = \delta_{ij} + u_{i,j}$

$$x(X, t) \approx x(t=0) + \Delta t \frac{\partial x}{\partial t}$$

$\mathbb{R}^2$

$$F_{ij} = \frac{1}{2} (C_{ij} - \delta_{ij})$$

$$= \frac{1}{2} (F_{ik}^T F_{kj} - \delta_{ij})$$

$$= \frac{1}{2} \left[ \left( \delta_{ki} + \frac{\partial u_k}{\partial x_i} \right) \left( \delta_{kj} + \frac{\partial u_k}{\partial x_j} \right) - \delta_{ij} \right]$$

$$= \frac{1}{2} \left( \frac{\partial u_j}{\partial x_i} + \frac{\partial u_i}{\partial x_j} + \frac{\partial u_k}{\partial x_i} \frac{\partial u_k}{\partial x_j} \right)$$

$$F_{ij} = \delta_{ij} + \varepsilon u_{i,j}$$

$$B = F F^T$$

$$B^{-1} = F^T F^{-1}$$

$$F_{ij} F_{jk} = (\delta_{ik} + \varepsilon u_{i,k})$$

$$e_{ij} = \frac{1}{2} (\delta_{ij} - B_{ij}^{-1})$$

$$= \frac{1}{2} (\delta_{ij} - F_{ki}^{-1} F_{kj}^{-1})$$

$$= \frac{1}{2} \left( \delta_{ij} - \left( \delta_{ki} - \frac{\partial u_k}{\partial x_i} \right) \left( \delta_{kj} - \frac{\partial u_k}{\partial x_j} \right) \right)$$

$$= \frac{1}{2} \left( \frac{\partial u_j}{\partial x_i} + \frac{\partial u_i}{\partial x_j} - \frac{\partial u_k}{\partial x_i} \frac{\partial u_k}{\partial x_j} \right)$$

$$\frac{\partial u_i}{\partial x_j} = \eta \frac{\partial \tilde{u}_i}{\partial x_j}$$

$$= \eta \frac{\partial \tilde{u}_i}{\partial x_k} \frac{\partial x_k}{\partial x_j}$$

$$= \eta \frac{\partial \tilde{u}_i}{\partial x_k} \left( \delta_{kj} + \eta \frac{\partial \tilde{u}_k}{\partial x_j} \right)$$

$$= \eta \frac{\partial \tilde{u}_i}{\partial x_j} + \eta^2 \frac{\partial \tilde{u}_i}{\partial x_k} \frac{\partial \tilde{u}_k}{\partial x_j} = \frac{\partial \tilde{u}_i}{\partial x_j}$$