

Lecture 7: Maximum Normal & Shear stresses

Logistics: - HW3 due Thursday

Last time: - Orthogonal tensors $\underline{\underline{Q}}^T \underline{\underline{Q}} = \underline{\underline{I}}$ $\det(\underline{\underline{Q}}) = \pm 1$
rotation or reflection

- Change in basis: $\{\underline{\underline{e}}_i\}$ $\{\underline{\underline{e}}'_i\}$;

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

$\Rightarrow \underline{\underline{A}}$ rotation

$$\underline{\underline{v}} = \underline{\underline{A}} \underline{\underline{v}}' \quad \underline{\underline{S}} = \underline{\underline{A}} \underline{\underline{S}}' \underline{\underline{A}}^T$$

- Invariance of $\text{tr}[\underline{\underline{S}}]$ & $\det[\underline{\underline{S}}]$

- Eigen problem: $\underline{\underline{S}} \underline{\underline{v}} = \lambda \underline{\underline{v}}$

$\underline{\underline{S}}$ is sym. pos. def.

$\Rightarrow \underline{\underline{v}}$'s are orthogonal

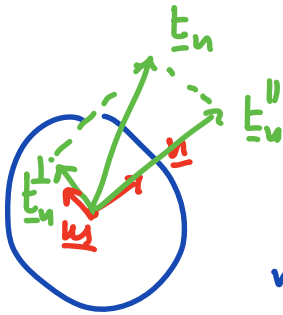
- Spectral decomposition

$$\underline{\underline{S}} = \sum_{i=1}^3 \lambda_i \underline{\underline{v}}_i \otimes \underline{\underline{v}}_i$$

- Principle invariants

Today: Max Normal & Shear stresses

Normal and Shear Stress



$$\underline{t}_n = \underline{\underline{\sigma}} \underline{n}$$

$$\underline{P}'' = \underline{n} \otimes \underline{n} \quad \underline{P}^\perp = \underline{I} - \underline{n} \otimes \underline{n} = \underline{m} \otimes \underline{m}$$

$$\text{normal stress} = \underline{t}_n'' = \underline{P}'' \underline{t}_n = \sigma_n \underline{n}$$

$$\text{shear stress} : \underline{t}_n^\perp = \underline{P}^\perp \underline{t}_n = \tau \underline{m}$$

$$\sigma_n = \underline{n} \cdot \underline{t}_n = \underline{n} \cdot \underline{\underline{\sigma}} \underline{n}$$

or

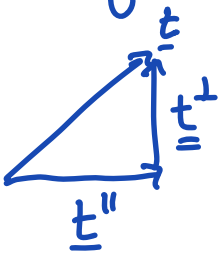
$$\sigma_n = n_i \sigma_{ij} n_j$$

$$\tau = \underline{m} \cdot \underline{t}_n = \underline{m} \cdot \underline{\underline{\sigma}} \underline{n}$$

or

$$\tau = m_i \sigma_{ij} n_j$$

From geometry:



$$\underline{t} = \underline{t}'' + \underline{t}^\perp$$

$$|\underline{t}|^2 = |\underline{t}''|^2 + |\underline{t}^\perp|^2$$

$$= |\sigma_n \underline{n}|^2 + |\tau \underline{m}|^2 =$$

$$|\underline{t}|^2 = \sigma_n^2 + \tau^2$$

Extremal stress values

→ important for failure

We all know this is related to eigenvalues

But why?

I, Extremal normal stresses

Given any $\underline{\underline{\sigma}}$ at \underline{x} what are

the unit normal \underline{n} corresponding to

extrema in σ_n ?

Constrained optimization problem

find extrema of $\sigma_n = \sigma_n(\underline{n}) = \underline{n} \cdot \underline{\underline{\sigma}} \underline{n}$

with constraint that $|\underline{n}| = 1$ $|\underline{n}|^2 = \underline{n} \cdot \underline{n} = 1$

$$\underline{n} \cdot \underline{n} - 1 = g(\underline{n})$$

⇒ Lagrange multiplier method

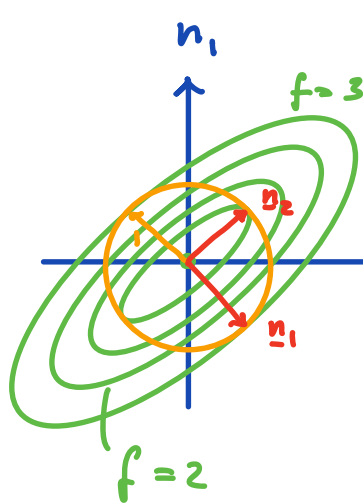
$$\mathcal{L}(\underline{n}, \lambda) = \underline{n} \cdot \underline{\underline{\delta}} \underline{n} - \lambda (\underline{n} \cdot \underline{n} - 1)$$

$$\mathcal{L}(n_i, \lambda) = \underbrace{n_i \delta_{ij} n_j}_{\text{function to optimize}} - \lambda \underbrace{(n_i n_i - 1)}_{\text{constraint}}$$

↑
Lagrange multiplier

Function $f(\underline{u}) = \underline{u} \cdot \underline{\underline{\delta}} \underline{u}$ is quadratic

if $\lambda_i > 0 \rightarrow$ level sets are ellipses



$$\underline{n} = \begin{pmatrix} n_1 \\ n_2 \end{pmatrix}$$

$$f(\underline{u}) = \underline{0} \cdot \underline{\underline{\delta}} \underline{0} = \underline{0}$$

$$f(\underline{u}) = \underline{\delta}_n(\underline{u}) = \underline{u} \cdot \underline{\underline{\delta}} \underline{u} -$$

unconstrained minimum

$$\underline{n} = \underline{0} \quad f(\underline{0}) = 0$$

Method of Lagrange multipliers

→ extremal values are the stationary points of

$$\mathcal{L}(n_i, \lambda) = n_i \delta_{ij} n_j - \lambda (n_i n_i - 1)$$

$$\frac{\partial \mathcal{L}}{\partial \lambda} = n_i n_i - 1 = 0$$

$$\frac{\partial \mathcal{L}}{\partial n_k} = \sigma_{ij} \left(\frac{\partial n_i}{\partial n_k} n_j + n_i \frac{\partial n_j}{\partial n_k} \right) - \lambda (2 n_i \frac{\partial n_i}{\partial n_k}) = 0$$

common notation $\frac{\partial n_i}{\partial n_k} = n_{i,k}$

$$\underline{n} = \begin{pmatrix} n_1 \\ n_2 \\ n_3 \end{pmatrix} \quad \frac{\partial n_1}{\partial n_1} = 1 \quad \frac{\partial n_1}{\partial n_3} = 0 \quad \frac{\partial n_1}{\partial n_2} = 0$$

$$\Rightarrow n_{i,k} = \delta_{ik} \quad n_{j,k} = \delta_{jk}$$

$$\frac{\partial \mathcal{L}}{\partial n_k} = \sigma_{ij} (\delta_{ik} n_j - n_i \delta_{jk}) - \lambda (2 n_i \delta_{ik}) = 0$$

$$= \sigma_{kj} n_j - \sigma_{ik} n_i - \lambda (2 n_k) = 0$$

$$\underline{\sigma} = \underline{\sigma}^T \quad \sigma_{kj} = \sigma_{jk}$$

$$= \sigma_{jk} n_j - \sigma_{ik} n_i - 2 \lambda n_k$$

rename dummy $j \rightarrow i$

$$= 2 (\sigma_{ik} n_k - \lambda n_k) = 0$$

switch to symbolic notation

$$\underline{\sigma} \underline{n} - \lambda \underline{n} = 0$$

$$\boxed{(\underline{\underline{\sigma}} - \lambda \underline{\underline{I}}) \underline{n} = 0}$$

Constrained optimization \rightarrow eigenvalue problem.

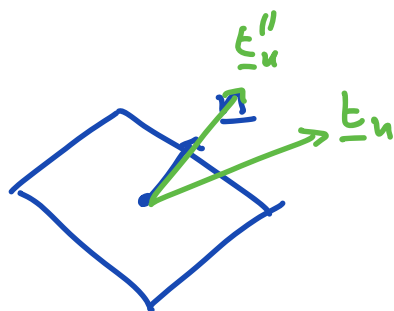
\Rightarrow eigen vectors of $\underline{\underline{\sigma}}$ are the directions of max & min normal stress.

To see that λ 's are associated normal stresses you dot eigen problem with \underline{n} .

$$\underline{n} \cdot (\underline{\underline{\sigma}} - \lambda \underline{\underline{I}}) \underline{n} = 0$$

$$\underbrace{\underline{n} \cdot \underline{\underline{\sigma}} \underline{n}} - \lambda \underline{n} \cdot \underline{n} =$$

$$\boxed{\sigma_n = \lambda}$$



$$\underline{t}_n = \underline{\underline{\sigma}} \underline{n}$$

$$\sigma_n = \underline{n} \cdot \underline{t}_n = \underline{n} \cdot \underline{\underline{\sigma}} \underline{n}$$

λ 's are principal stresses $\Rightarrow \lambda_i = \sigma_i$

$$\sigma_1 > \sigma_2 > \sigma_3$$

$\sigma_1 = \text{max normal stress}$

$\sigma_3 = \text{min normal stress}$

\underline{n}_i 's are the principal directions

What are tractions on principal planes
(planes associated \underline{n}_i 's)

$$\underline{t}_{\underline{n}_i} = \underline{\sigma} \underline{n}_i = \left(\sum_{j=1}^3 \sigma_j \underline{n}_j \otimes \underline{n}_j \right) \underline{n}_i = \sum_{j=1}^3 \sigma_j \underbrace{(\underline{n}_j \cdot \underline{n}_i)}_{\delta_{ij}} \underline{n}_j$$

$$\underline{t}_{\underline{n}_i} = \sigma_i \underline{n}_i$$

traction is equal to normal stress $\underline{t} \parallel \underline{n}_i$

\Rightarrow no shear stresses on principal planes.

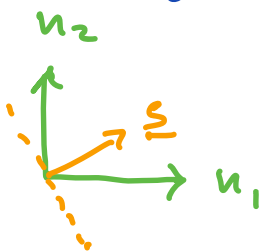
$$\underline{t}_{\underline{n}_i}^\perp = \underline{0}$$

II) Extremal shear stresses

Given the principal dir. $\{\underline{n}_i\}$ at \underline{x} what

is the direction/unit vector $\underline{s} = [s_1, s_2, s_3]^T$

that gives the max. & min. values of τ ?



$$\underline{s} = s_i \underline{n}_i \quad s_i = \underline{s} \cdot \underline{n}_i$$

traction in dir \underline{s}

$$\begin{aligned} \underline{t}_{\underline{s}} &= \underline{\sigma} \underline{s} = \left(\sum_{j=1}^3 \sigma_j \underline{n}_j \otimes \underline{n}_j \right) (s_j \underline{n}_j) \\ &= \sum_{j=1}^3 \sigma_j s_j (\underline{n}_j \otimes \underline{n}_j) \underline{n}_j \end{aligned}$$

only 1 index

$$= \sum_{i=1}^3 \sigma_i s_i \underbrace{\eta_i}_{(\eta_i = \eta_j)} \delta_{ij}$$

traction on \underline{s}

$$\underline{t}_s = \sigma_1 s_1 \underline{n}_1 + \sigma_2 s_2 \underline{n}_2 + \sigma_3 s_3 \underline{n}_3$$

need expression for shear stress τ

$$|\underline{t}_s|^2 = \sigma_n^2 + \tau^2 \Rightarrow \tau^2 = |\underline{t}_s|^2 - \sigma_n^2$$

normal stress: $\sigma_n = \underline{s} \cdot \underline{t}_s = \sigma_1 s_1^2 + \sigma_2 s_2^2 + \sigma_3 s_3^2$

$$\tau^2 = \underline{t}_s \cdot \underline{t}_s - \sigma_n^2 = \sigma_1^2 s_1^2 + \sigma_2^2 s_2^2 + \sigma_3^2 s_3^2 - (\sigma_1 s_1^2 + \sigma_2 s_2^2 + \sigma_3 s_3^2)^2$$

shear stress in principal frame:

$$\tau^2 = \sum_{i=1}^3 \sigma_i^2 s_i^2 - \left(\sum_{i=1}^3 \sigma_i s_i^2 \right)^2$$

$$\tau = \tau(\underbrace{\sigma_1, \sigma_2, \sigma_3}_{\text{given}}, \underbrace{s_1, s_2, s_3}_{\text{unknown}})$$

Looking for extrema of $\tau^2(\tau)$ under the

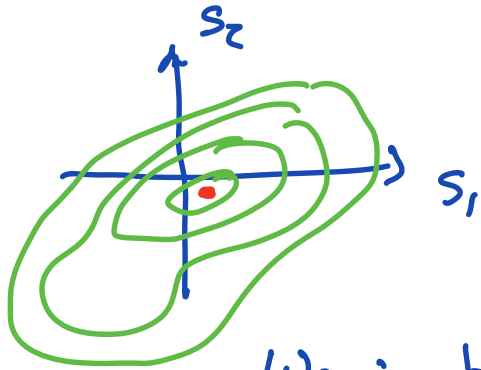
constraint that $|s|=1$ $s_1^2 + s_2^2 + s_3^2 = 1$

$$s_3^2 = 1 - s_1^2 - s_2^2$$

Solve constrained optimization problem by direct elimination:

eliminate s_3^2 by substituting $\underline{s_3^2 = 1 - s_1^2 - s_2^2}$

$$\Rightarrow \tau^2 = \tau^2(s_1, s_2)$$



because we are substituted constraint

\Rightarrow solve for max min directly

We just need to find

$$\frac{\partial \tau^2}{\partial s_1} = \frac{\partial \tau^2}{\partial s_2} = 0$$

$$\frac{\partial \tau^2}{\partial s_1} = 2s_1(\sigma_1 - \sigma_3) \left\{ \sigma_1 - \sigma_3 - 2[(\sigma_1 - \sigma_3)s_1^2 + (\sigma_2 - \sigma_3)s_2^2] \right\} = 0$$

$$\frac{\partial \tau^2}{\partial s_2} = 2s_2(\sigma_2 - \sigma_3) \left\{ \sigma_2 - \sigma_3 - 2[(\sigma_1 - \sigma_3)s_1^2 + (\sigma_2 - \sigma_3)s_2^2] \right\} = 0$$

First solution (trivial): $s_1 = s_2 = 0 \Rightarrow s_3 = 1$

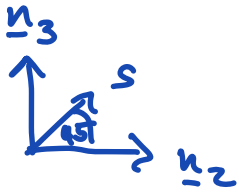
$$\Rightarrow \underline{s} = \pm \underline{n}_3$$

$$\tau^2 = 0$$

\Rightarrow minimum in shear stress

on principal planes

Second solution: $s_1 = 0$



$$\Rightarrow s_2 = \pm \frac{1}{\sqrt{2}}$$

$$s_3 = \pm \frac{1}{\sqrt{2}}$$

$$\underline{s} = \pm \frac{1}{\sqrt{2}} (\underline{n}_2 + \underline{n}_3)$$

shear stress: $\tau^2 = \left(\frac{\sigma_2 - \sigma_3}{2}\right)^2$ $\tau = \frac{1}{2}(\sigma_2 - \sigma_3)$

We have two solutions:

min: $\tau = 0$ for $\underline{s} = \pm \underline{n}_3$

max: $\tau = \frac{1}{2}(\sigma_2 - \sigma_3)$ for $\underline{s} = \frac{1}{\sqrt{2}}(\pm \underline{n}_2 \pm \underline{n}_3)$

two additional pairs of solus by
eliminating s_1 or s_2 in total

we have:

Min. shear stresses:

$$\tau = 0 \quad \text{on} \quad \underline{s} = \pm \underline{n}_1, \quad \underline{s} = \pm \underline{n}_2, \quad \underline{s} = \pm \underline{n}_3$$

Max. shear stresses

$$\tau_{23} = \frac{1}{2} (\sigma_2 - \sigma_3) \quad \text{on} \quad s_{23} = \frac{1}{\sqrt{2}} (\pm n_2 \pm n_3)$$

$$\tau_{13} = \frac{1}{2} (\sigma_1 - \sigma_3) \quad \text{on} \quad s_{13} = \frac{1}{\sqrt{2}} (\pm n_1 \pm n_3)$$

$$\tau_{12} = \frac{1}{2} (\sigma_1 - \sigma_2) \quad \text{on} \quad s_{12} = \frac{1}{\sqrt{2}} (\pm n_1 \pm n_2)$$

