Lecture 8: Normal and shear strem

Logistics: - no hw this week

- no office his unless requested

Last time: · Cauchy stress tensor t=on

· Equilibrium Equations

$$\vec{L} = 0 \quad \Rightarrow \quad \triangle \cdot \vec{e} + b \vec{p} = 0$$

6 equs but 9 unknowns!

Today: - Normal & shear stress

- principal strenes
- Stren ellipsoid
- Simple states of stress

Normal & Shew shorres

Projection matrices:

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normal stres: t" = P"t" = (n.t") n = 6" "

shear shess: th = Pt tn = (m.tn) m = T m

Hag nitudes

From geometry:
$$\underline{t}_n = \underline{t}_n'' + \underline{t}_n'$$

Extremal stress values

Given = at x what muit normals no corresponding to min & max normal stress on

⇒ constraind ophimization problem

need to Sind meex and min = ou(n)

but with constraint 111=1

Lagrange multiplier method

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

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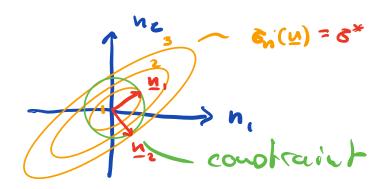
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= n; 8; n; - x (n;n; - 1)

Ter visualization assume & is spot (compromis)

6 = 10 & 10 > is quadratic



To find extremal values we used to find stationary points of $2(\underline{n}, \lambda)$ $L = n; \delta; n; -\lambda (n; n; -1)$

 $\frac{2x}{2x} = n_1 n_1 - 1 = 0 \Rightarrow (n_1 + n_2 n_3 + n_3 n_3 + n_4 n_3 n_4) - \lambda (n_1 + n_2 n_3 n_4) - \lambda (n_1 + n_2 n_3 n_4) + \delta_{1k} n_2 + \delta_{1k} n_3 + \delta_{1k} n_3$

 $= \sigma_{kj} u_j + \sigma_{ik} u_i - 2\lambda u_k$ $= 2(\sigma_{ik} u_k - \lambda u_k) = 0$

lu symbolic notation

 $\sqrt{2} \vec{n} - y \vec{n} = 0 \quad (|\vec{n}| = 1)$

=> (\vec{c} - \lambde{I}) \vec{n} =0 eigenvalue problem

Clearly n; directions to max, win

What is λ ? $\underline{n} \cdot (\underline{\delta} - \lambda \underline{I}) \underline{n} = 0$ $\underline{n} \cdot (\underline{\delta} - \lambda \underline{I}) \underline{n} = 0$ $\underline{\delta}_{n} = \lambda$

λ; s are the principal normal stresses λ;=ε;
n;'s are the principal directions of <u>σ</u>

Since === all e; s are real and
the { v; } forms on orthonormal basis

=> == = 5 o; v; & v;

Tractions ia principal dir.

 $\frac{\mathbf{f}^{n!}}{\mathbf{f}^{n!}} = \frac{\mathbf{g}^{n}}{\mathbf{g}^{n}} = \mathbf{g}^{n} \cdot \mathbf{g}^{n}$ $\mathbf{f}^{n} = \mathbf{g}^{n} \cdot \mathbf{g$

M.

If ei's ar distinct e, > 02 703 then o, and o, ore man a min normal stronger.

Hax & min shear stresses

Given principal dir Enis at x
what is unit vector $S = [S_1, S_2, S_3]$ Hat gives max & min values of sheer shew?

In frame { ui} to is

to = g = (\frac{2}{5} c; u; \omega u_i) s

= 6, 8, \frac{1}{12} c \frac{1}{2} c \frac{1}{2

mag un tuols of shew on \underline{s} $\begin{bmatrix}
 \xi_1 & \xi_2 \\
 \xi_2 & \xi_3
 \end{bmatrix}
 \begin{bmatrix}
 \xi_1 & \xi_2 \\
 \xi_3 & \xi_3
 \end{bmatrix}
 = <math>\underline{s} \cdot \underline{t}_s = \underline{s}_1 \cdot \underline{s}_1 + \underline{s}_2 \cdot \underline{s}_2^2 + \underline{s}_3 \cdot \underline{s}_3^2 = \underline{s}_1 \cdot \underline{s}_1^2
 \end{bmatrix}$

index notation

$$T^{2} = \sum_{i=1}^{5} G_{i}^{2} s_{i}^{2} - \left(\sum_{i=1}^{5} G_{i} S_{i}^{2}\right)^{2}$$

shear shers ou some plane with normal

s in { n; }

looking for extreme 2 ander constraint

ets messer messer

Hue we choose direct elimination.

Eliminate
$$s_{1}^{2} = 1$$
 $s_{1}^{2} + s_{2}^{2} + s_{3}^{2} = 1$
 $s_{2}^{2} = 1 - s_{2}^{2} - s_{1}^{2}$
 $\Rightarrow T^{2} = T^{2}(s_{1}, s_{2})$

$$\frac{\partial C}{\partial s_{1}} = 2 \frac{s_{1}}{s_{2}} \left[(s_{1} - s_{3}) \left\{ s_{1} - s_{2} - 2 \left[(s_{1} - s_{3}) s_{1}^{2} + (s_{2} - s_{3}) s_{2}^{2} \right] \right\} = 0$$

$$\frac{\partial C}{\partial s_{2}} = 2 \frac{s_{2}}{s_{2}} \left[(s_{2} - s_{3}) \left\{ s_{2} - s_{3} - 2 \left[(s_{1} - s_{3}) s_{2}^{2} + (s_{2} - s_{3}) s_{2}^{2} \right] \right\} = 0$$

First solution: $s_1 = s_2 = 0 \Rightarrow s_3 = \pm 1 \Rightarrow \leq = \pm n_3$ $T^2 = \sigma_3^2 \cdot 1 - (\sigma_3 \cdot 1)^2 = 0$ $\Rightarrow \text{ minimum in shear shows}$

shear shess vouilles on principal planes and logously this can be shown for M,, M2

Second solution: 5, -0

$$\frac{\partial z^{2}}{\partial s_{e}} = (\epsilon_{z} - \epsilon_{s}) - 2[(\epsilon_{c} - \epsilon_{s}) s_{z}^{2}] = 0$$

$$(\epsilon_{z} - \epsilon_{s}) (1 - 2 s_{z}^{2}) = 0 \Rightarrow s_{e} = \pm \frac{1}{|z|}$$

$$\int_{0}^{\infty} x^{2} + s_{e}^{2} + s_{s}^{2} = 1 \Rightarrow s_{s} = \pm \frac{1}{|z|}$$

$$s_{e} = \pm \frac{1}{|z|} n_{z} \pm \frac{1}{|z|} n_{3} = \pm \frac{1}{|z|} (n_{z} + n_{s})$$

We have following two solutions
min.: T=0 for $\underline{s}=\pm \underline{n}_{s}$

max.: $T = \frac{1}{2} (\delta_e - \delta_3)$ for $s = \frac{1}{2} \frac{\pi_e}{\sqrt{2}} + \frac{\pi_3}{\sqrt{2}}$

=> similarly we can find two additional pairs of solus by eliminating s, and s, and s, and following similar steps

Muinam shear stresses

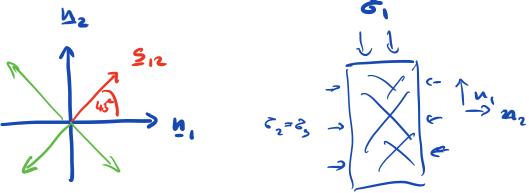
Maximum shear shess

$$T_{23} = \frac{1}{2} (G_2 - G_3) \quad \text{ou} \quad \underline{S}_{23} = \frac{1}{\sqrt{2}} (\pm \underline{N}_2 \pm \underline{N}_3)$$

$$T_{13} = \frac{1}{2} (G_1 - G_3) \quad \text{ou} \quad \underline{S}_{13} = \frac{1}{\sqrt{2}} (\pm \underline{N}_1 \pm \underline{N}_3)$$

$$T_{12} = \frac{1}{2} (G_1 - G_2) \quad \text{ou} \quad \underline{S}_{12} = \frac{1}{\sqrt{2}} (\pm \underline{N}_1 \pm \underline{N}_2)$$

assume 6, 2 6, 2 6,



=> sets of conjugate shear fractures at 45° to 5,