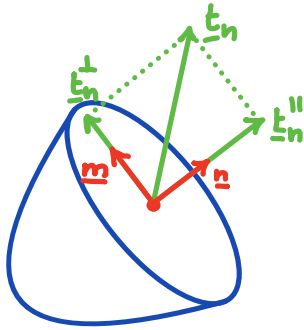


Normal and Shear Stresses



Consider an arbitrary surface in B with normal \underline{n} . Then we have the two projection matrices

$$\underline{P}'' = \underline{n} \otimes \underline{n} \quad \text{and} \quad \underline{P}' = \underline{I} - \underline{n} \otimes \underline{n} = \underline{m} \otimes \underline{m}$$

that define the

$$\text{normal stress: } \underline{t}_n'' = \underline{P}'' \underline{t}_n = (\underline{n} \cdot \underline{t}_n) \underline{n} = \sigma_n \underline{n}$$

$$\text{shear stress: } \underline{t}_n' = \underline{P}' \underline{t}_n = (\underline{m} \cdot \underline{t}_n) \underline{m} = \tau \underline{m}$$

The magnitudes of these stresses are

$$\sigma_n = \underline{n} \cdot \underline{t}_n = \underline{n} \cdot \underline{\sigma} \underline{n} \quad \text{or} \quad \sigma_n = n_i \sigma_{ij} n_j$$

$$\tau = \underline{m} \cdot \underline{t}_n = \underline{m} \cdot \underline{\sigma} \underline{n} \quad \text{or} \quad \tau = m_i \sigma_{ij} n_j$$

If $\sigma_n > 0$ the normal stresses are tensile if $\sigma_n < 0$ the normal stresses are compressive.

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

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

Extremal Stress Values

I, Maximum and Minimum Normal Stresses

Given a state of stress $\underline{\underline{\sigma}}$ at point \underline{x} , what are the unit normals \underline{n} corresponding to min. and max. normal stress σ_n .

This is a constrained optimization problem, because we want to find extrema of the function $\sigma_n = \sigma_n(\underline{n})$ with the constraint that $|\underline{n}| = 1$.

Lagrange multiplier method

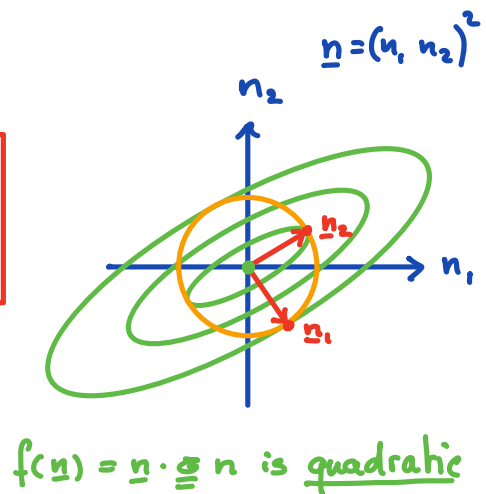
$$\mathcal{L}(\underline{n}, \lambda) = \underline{n} \cdot \underline{\underline{\sigma}} \underline{n} - \lambda (\underline{n} \cdot \underline{n} - 1)$$

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

function

constraint

Lagrange multiplier



Function $f(\underline{n}) = \underline{n} \cdot \underline{\underline{\sigma}} \underline{n}$ is quadratic in components of \underline{n} .

If eigenvalues of $\underline{\underline{\sigma}}$ are positive then the level sets of $f(\underline{n})$ are ellipsoids as shown.

The extremal values are the stationary points of $\mathcal{L}(\underline{n}, \lambda)$

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

$$\frac{\partial \mathcal{L}}{\partial n_k} = \sigma_{ij} (n_{i,k} n_j + n_i n_{j,k}) - \lambda (2 n_i n_{i,k}) = 0$$

$$\text{where } n_{i,k} = \delta_{i,k} \quad n_{j,k} = \delta_{j,k}$$

$$= \sigma_{ij} (\delta_{i,k} n_j + \delta_{j,k} n_i) - \lambda (2 n_i \delta_{i,k})$$

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

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

In symbolic notation: $(\underline{\underline{\sigma}} - \lambda \underline{\underline{I}}) \underline{n} = 0$ and $|\underline{n}| = 1$

The Lagrange multiplier method leads to an eigen problem, where the Lagrange multiplier, λ , is the eigenvalue and the normal, \underline{n} , the eigenvector.

We can see that λ is the magnitude of the normal stress by taking the dot product of eigenproblem with \underline{n} .

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

Hence to find the extremal stress values we must find the eigenvalues λ_i and eigenvectors \underline{n}_i .

λ_i 's are the principal normal stresses $\Rightarrow \lambda_i = \sigma_i$

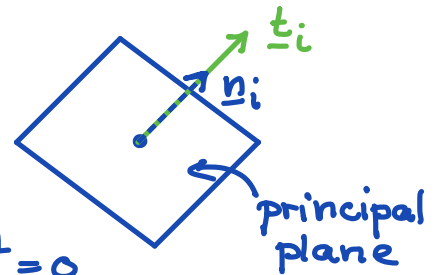
\underline{n}_i 's are the principal directions of $\underline{\underline{\sigma}}$

Since $\underline{\underline{\sigma}} = \underline{\underline{\sigma}}^T$ all λ_i are real and the set $\{\underline{n}_i\}$ form a mutually orthogonal basis, so that $\underline{\underline{\sigma}}$ can be represented as $\underline{\underline{\sigma}} = \sum_{i=1}^3 \sigma_i \underline{n}_i \otimes \underline{n}_i$

The tractions in the principal directions are

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

Since $\underline{t}_i \parallel \underline{n}_i$ there is no shear stress on the principal planes, $\underline{t}_i^\perp = 0$.



If the σ_i 's are distinct and ordered $\sigma_1 > \sigma_2 > \sigma_3$

then σ_1 and σ_3 are the max. and min. normal stresses.

II. Maximum and minimum shear stresses

Given the principal directions $\underline{n}_1, \underline{n}_2$ and \underline{n}_3 at x what is the unit vector $\underline{s} = [s_1 \ s_2 \ s_3]$ that gives the max. and min. values of the shear stresses τ ?

In the frame of the principal directions $\{\underline{n}_i\}$ the traction vector associated with \underline{s} is

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

The magnitudes of normal, σ_n , and shear stress, τ , are

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

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

Hence we have the following expression for the shear stress $\tau^2 = \sum_{i=1}^3 \sigma_i^2 s_i^2 - \left(\sum_{i=1}^3 \sigma_i s_i^2 \right)^2$ we are looking for the extremal values of τ^2 under the constraint $|\underline{s}|^2 - 1 = 0$.

\Rightarrow Solve using Lagrange mult. or direct elimination.

I) Eliminate $s_3^2 = 1 - s_1^2 - s_2^2 \Rightarrow \tau^2 = \tau^2(s_1, s_2)$.

We just need to find $\frac{\partial \tau^2}{\partial s_1} = \frac{\partial \tau^2}{\partial s_2} = 0$.

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

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

First solution: $s_1 = s_2 = 0 \Rightarrow s_3 = 1 \quad \underline{s} = \pm \underline{n}_3$

$$\tau^2 = \sigma_3^2 \cdot 1 - (\sigma_3 \cdot 1)^2 = 0$$

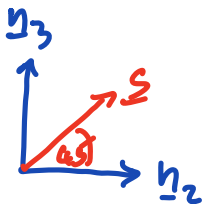
\Rightarrow minimum in the shear stress

which vanishes on principal plane

Second solution: $s_1 = 0$

$$\frac{\partial \tau^2}{\partial s_2} = \sigma_2 - \sigma_3 - 2[(\sigma_2 - \sigma_3)s_2^2] = 0$$

$$(\sigma_2 - \sigma_3)(1 - 2s_2^2) = 0 \Rightarrow s_2 = \pm \frac{1}{\sqrt{2}}$$



$$\text{from } s_2^2 + s_3^2 = 1 \Rightarrow s_3 = \pm \frac{1}{\sqrt{2}}$$

$$\Rightarrow \underline{s} = \pm \frac{1}{\sqrt{2}} \underline{n}_2 \pm \frac{1}{\sqrt{2}} \underline{n}_3$$

$$\tau^2 = \frac{\sigma_1^2}{2} + \frac{\sigma_2^2}{2} - \left(\frac{\sigma_1}{2} + \frac{\sigma_2}{2} \right)^2$$

$$= \frac{\sigma_1^2}{2} + \frac{\sigma_2^2}{2} - \left(\frac{\sigma_1^2}{4} + 2\frac{\sigma_1}{2}\frac{\sigma_2}{2} + \frac{\sigma_2^2}{4} \right)$$

$$\tau^2 = \left(\frac{\sigma_2}{2}\right)^2 - 2 \frac{\sigma_2}{2} \frac{\sigma_3}{2} + \left(\frac{\sigma_3}{2}\right)^2 = \left(\frac{\sigma_2 - \sigma_3}{2}\right)^2$$

We have the following two solutions:

$$\begin{aligned} \text{min. } \tau &= 0 & \text{for } \underline{s} &= \pm \underline{n}_3 \\ \text{max. } \tau &= \frac{1}{2}(\sigma_2 - \sigma_3) & \text{for } \underline{s} &= \pm \frac{\underline{n}_2}{\sqrt{2}} \pm \frac{\underline{n}_3}{\sqrt{2}} \end{aligned}$$

Two additional pairs of solutions can be obtained by eliminating \underline{n}_1 or \underline{n}_2 from τ^2 and following similar steps. So that we have

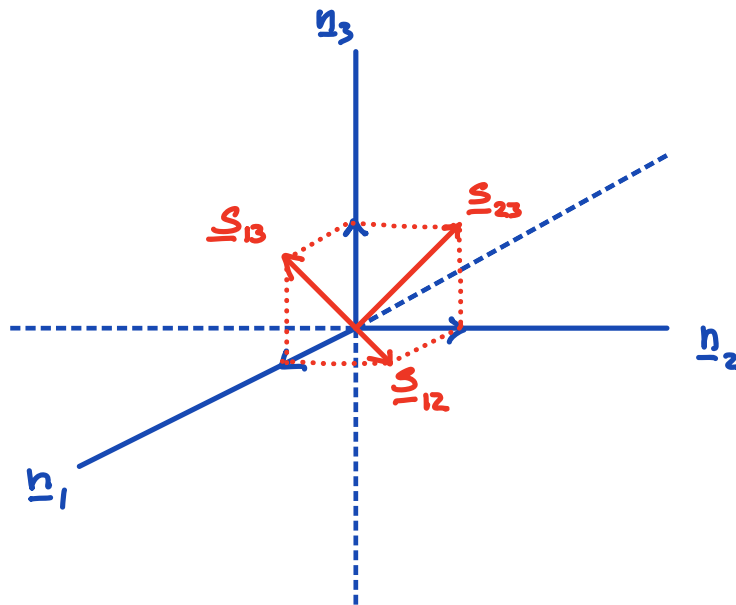
Minimum shear stresses:

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

Maximum shear stresses:

$$\begin{aligned} \tau_{23} &= \frac{1}{2}(\sigma_2 - \sigma_3) & \text{on} & \quad \underline{s}_{23} = \frac{1}{\sqrt{2}}(\pm \underline{n}_2 \pm \underline{n}_3) \\ \tau_{13} &= \frac{1}{2}(\sigma_1 - \sigma_3) & \text{on} & \quad \underline{s}_{13} = \frac{1}{\sqrt{2}}(\pm \underline{n}_1 \pm \underline{n}_3) \\ \tau_{12} &= \frac{1}{2}(\sigma_1 - \sigma_2) & \text{on} & \quad \underline{s}_{12} = \frac{1}{\sqrt{2}}(\pm \underline{n}_1 \pm \underline{n}_2) \end{aligned}$$

where we assume $\sigma_1 \geq \sigma_2 \geq \sigma_3$



Note: G&S do this with Lagrange multipliers but it leads to odd expressions in index notation, such as

$$4 \left(\sum_{j=1}^3 n_j^2 \sigma_j \right) n_i \delta_i = 2\lambda n_i \quad ?$$

where 'i' seems to be a dummy on the l.h.s. but a free index on the r.h.s.

\Rightarrow we did it the pedestrian way.