

# assignment1\_solution

March 11, 2025

## 1 Homework Assignment 1

**Note:** This is a computable IPython notebook who's source code can be downloaded [here](#).

### 1.1 Problem 1

The motion of a certain continuous medium is defined by the equations

$$\begin{aligned}x_1 &= \frac{1}{2}(X_1 + X_2)e^t + \frac{1}{2}(X_1 - X_2)e^{-t}, \\x_2 &= \frac{1}{2}(X_1 + X_2)e^t - \frac{1}{2}(X_1 - X_2)e^{-t}, \\x_3 &= X_3\end{aligned}$$

1. Compute the following

1. The Green-Lagrange strain tensor  $E$
2. The linear (small) strain tensor  $\varepsilon$

Plot the 11, 22, and 12 components of  $E$  and  $\varepsilon$  on the same figure from time  $t = 0$  to  $t = 0.05$ .

2. Compute the following

1. The rate-of-deformation tensor  $D$
2. The rate-of-change of the small strain tensor  $\dot{\varepsilon} = \frac{d\varepsilon}{dt}$

Plot the 11, 22, and 12 components of  $D$  and  $\dot{\varepsilon}$  on the same figure from time  $t = 0$  to  $t = 0.05$ .

### Solution

This cell loads some important packages from sympy, numpy, and matplotlib that I will use the perform calculations and display the results neatly. In order to run this notebook, you will have to have these packages installed in your Python distribution.

```
[1]: from sympy import *
from sympy.matrices import *
import sympy.mpmath
from sympy.utilities.lambdify import lambdify
init_printing()

import numpy
```

```
%matplotlib inline
import matplotlib.pyplot as plt
#plt.style.available
#plt.style.use('bmh')
```

Defining which variables will be “symbolic” in nature, i.e., they will not take on numerical values.

```
[2]: t, X1, X2, X3 = symbols('t, X_1, X_2, X_3')
```

This defines the deformation mapping as in the problem statement.

```
[3]: x1 = Rational(1, 2) * (X1 + X2) * exp(t) + Rational(1, 2) * (X1 - X2) * exp(-t)
x2 = Rational(1, 2) * (X1 + X2) * exp(t) - Rational(1, 2) * (X1 - X2) * exp(-t)
x3 = X3
```

Now we compute the deformation gradient.

```
[4]: F = simplify(Matrix([[diff(x1, X1), diff(x1, X2), diff(x1, X3)],
                           [diff(x2, X1), diff(x2, X2), diff(x2, X3)],
                           [diff(x3, X1), diff(x3, X2), diff(x3, X3)]])); F
```

```
[4]:
```

$$\begin{bmatrix} \cosh(t) & \sinh(t) & 0 \\ \sinh(t) & \cosh(t) & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

And the Green-Lagrange strain

```
[5]: E = Rational(1, 2) * (F.T * F - eye(3)); simplify(E)
```

```
[5]:
```

$$\begin{bmatrix} \sinh^2(t) & \frac{1}{2} \sinh(2t) & 0 \\ \frac{1}{2} \sinh(2t) & \sinh^2(t) & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

We can immediately evaluate the linear “small” strain as well, directly with the deformation gradient.

```
[6]: epsilon = simplify(Rational(1,2) * (F.T + F) - eye(3)); simplify(epsilon)
```

```
[6]:
```

$$\begin{bmatrix} \cosh(t) - 1 & \sinh(t) & 0 \\ \sinh(t) & \cosh(t) - 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

This turns our symbolic output into actual functions of  $t$  that we can evaluate and plot

```
[7]: E11_function = lambdify(t, E[0,0], "numpy")
epsilon11_function = lambdify(t, epsilon[0,0], "numpy")

E22_function = lambdify(t, E[1,1], "numpy")
```

```

epsilon22_function = lambdify(t, epsilon[1,1], "numpy")

E12_function = lambdify(t, E[0,1], "numpy")
epsilon12_function = lambdify(t, epsilon[0,1], "numpy")

```

Evaluating the functions

```

[9]: t0 = numpy.linspace(0.0,0.05)

E11 = E11_function(t0)
epsilon11 = epsilon11_function(t0)

E22 = E22_function(t0)
epsilon22 = epsilon22_function(t0)

E12 = E12_function(t0)
epsilon12 = epsilon12_function(t0)

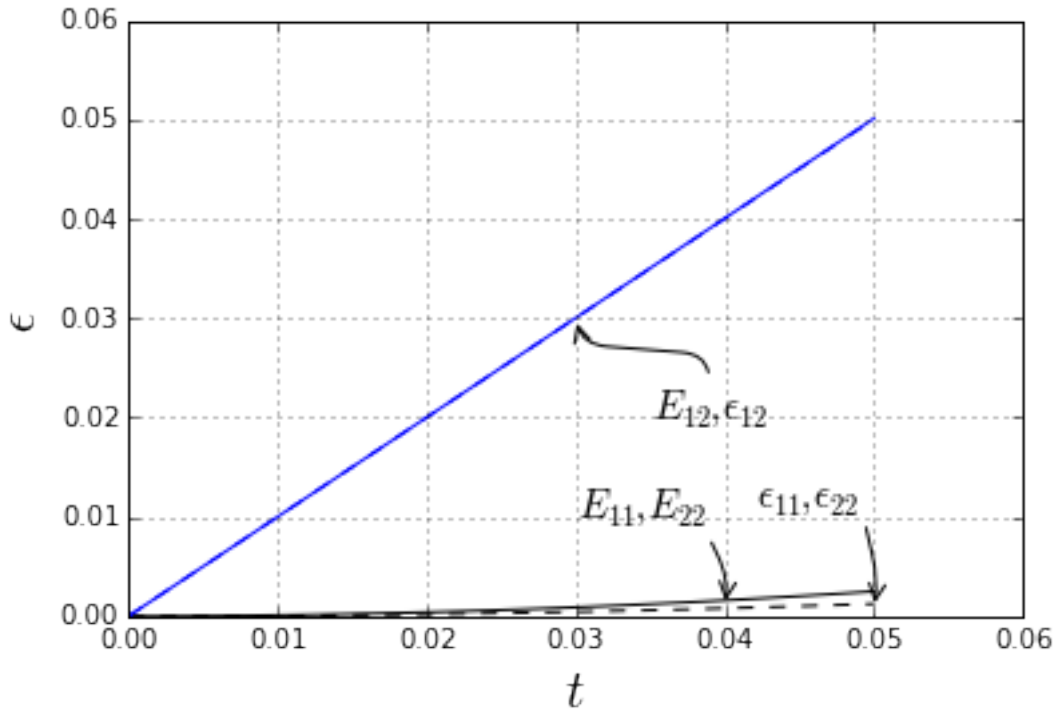
```

Plotting the results

```

[10]: fig = plt.figure(1)
ax = fig.add_subplot(111,xlabel='$t$', ylabel='$\epsilon$')
ax.plot(t0, E22,'k-', t0, epsilon22, 'k--',
        t0, E12,'b-', t0, epsilon12, 'b--');
plt.grid()
ax.annotate('$E_{11}$', xy=(0.04, 0.001), xycoords='data',
            xytext=(0.03, 0.01), textcoords='data',
            arrowprops=dict(arrowstyle="->",
            ↪connectionstyle="arc,angleA=0,armA=20,angleB=80,armB=10,rad=10"),
            fontsize='15'
            )
ax.annotate('$\epsilon_{11}$', xy=(0.05, 0.0008),
            ↪xycoords='data',
            xytext=(0.042, 0.011), textcoords='data',
            arrowprops=dict(arrowstyle="->",
            ↪connectionstyle="arc,angleA=0,armA=20,angleB=80,armB=10,rad=10"),
            fontsize='15'
            )
ax.annotate('$E_{12}$', xy=(0.03, 0.03), xycoords='data',
            xytext=(0.035, 0.02), textcoords='data',
            arrowprops=dict(arrowstyle="->",
            ↪connectionstyle="arc,angleA=90,armA=20,angleB=-80,armB=10,rad=10"),
            fontsize='15'
            );
ax.xaxis.label.set_size(20)
ax.yaxis.label.set_size(20)

```



Now compute the rate-of-deformation tensor, i.e. the symmetric part of the velocity gradient

```
[11]: Fdot = F.diff(t);
      L = expand(Fdot * F.inv());
      D = simplify(Rational(1,2) * (L.T + L)); D
```

[11]:

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

And the linear “small” strain-rate,  $\dot{\epsilon}$

```
[12]: epsilon_dot = epsilon.diff(t); epsilon_dot
```

[12]:

$$\begin{bmatrix} \sinh(t) & \cosh(t) & 0 \\ \cosh(t) & \sinh(t) & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Now we will turn the symbolic small strain components into functions that we can evaluate in time. It’s not necessary to perform this operation on the rate-of-deformation tensor because it has constant components

```
[13]: epsilon11_dot_function = lambdify(t, epsilon_dot[0,0], "numpy")

epsilon22_dot_function = lambdify(t, epsilon_dot[1,1], "numpy")

epsilon12_dot_function = lambdify(t, epsilon_dot[0,1], "numpy")
```

Evaluating the functions

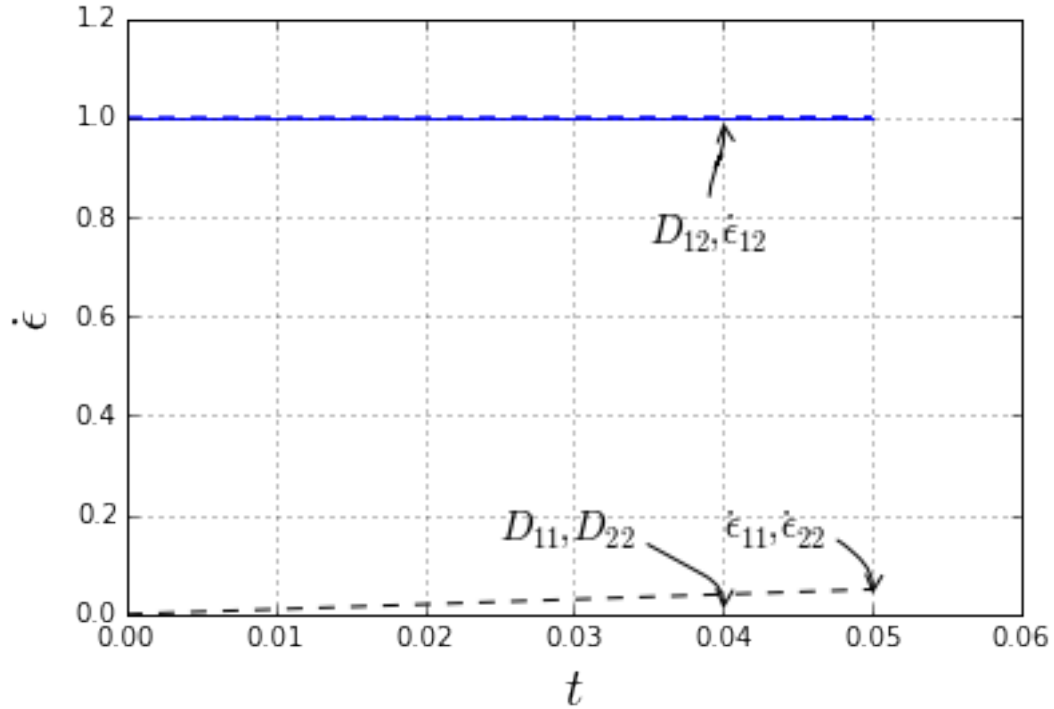
```
[14]: D11 = numpy.zeros_like(t0)
epsilon11_dot = epsilon11_dot_function(t0)

D22 = D11
epsilon22_dot = epsilon22_dot_function(t0)

D12 = numpy.ones_like(t0)
epsilon12_dot = epsilon12_dot_function(t0)
```

Plotting the results

```
[15]: fig = plt.figure(2)
ax = fig.add_subplot(111,xlabel='$t$', ylabel='$\dot{\epsilon}$')
ax.plot(t0, D11,'k-', t0, epsilon11_dot, 'k--',
        t0, D12,'b-', t0, epsilon12_dot, 'b--');
plt.grid()
ax.annotate('$D_{11}$, $D_{22}$', xy=(0.04, 0.0), xycoords='data',
           xytext=(0.025, 0.15), textcoords='data',
           arrowprops=dict(arrowstyle="->",
           ↪connectionstyle="arc,angleA=0,armA=20,angleB=80,armB=10,rad=10"),
           fontsize='15'
           )
ax.annotate('$\dot{\epsilon}_{11}$, $\dot{\epsilon}_{22}$', xy=(0.05, 0.03), ↪
           ↪xycoords='data',
           xytext=(0.04, 0.15), textcoords='data',
           arrowprops=dict(arrowstyle="->",
           ↪connectionstyle="arc,angleA=0,armA=20,angleB=80,armB=10,rad=10"),
           fontsize='15'
           )
ax.annotate('$D_{12}$, $\dot{\epsilon}_{12}$', xy=(0.04, 1.0), xycoords='data',
           xytext=(0.035, 0.75), textcoords='data',
           arrowprops=dict(arrowstyle="->",
           ↪connectionstyle="arc,angleA=90,armA=20,angleB=-80,armB=10,rad=10"),
           fontsize='15'
           );
ax.xaxis.label.set_size(20)
ax.yaxis.label.set_size(20)
```



## 1.2 Problem 2

Given the following stress tensor

$$\sigma = \begin{bmatrix} 36 & 27 & 0 \\ 27 & -36 & 0 \\ 0 & 0 & 18 \end{bmatrix}$$

Find:

1. the components of the traction vector acting on a plane with unit normal vector  $\hat{n}^T = [2/3, -2/3, 1/3]$
2. the magnitude of the traction vector found in (a)
3. its component in the direction of the normal
4. a. the angle between the traction vector and the normal

### Solution

Defining the stress tensor and normal vector

```
[16]: sigma = Matrix([[36, 27, 0],[27, -36, 0],[0, 0, 18]])
      n = Matrix([Rational(2, 3), Rational(-2, 3), Rational(1, 3)])
```

The traction vector is then  $\vec{t} = \sigma^T \hat{n}$  according to the Cauchy stress equation.

```
[17]: t = sigma.T * n; t
```

```
[17]:
```

$$\begin{bmatrix} 6 \\ 42 \\ 6 \end{bmatrix}$$

Computing the magnitude

```
[18]: sympy.mpmath.mp.pretty = True
      magnitude = sympy.mpmath.norm(t,2); magnitude
```

```
[18]: 42.8485705712571
```

And the projection in the direction of the normal.

```
[19]: t.T * n
```

```
[19]:
```

$$[-22]$$

The angle between the normal and the traction vector (in radians)

```
[20]: acos(((t / magnitude).T * n)[0])
```

```
[20]:
```

2.10998044394962

or in degree

```
[21]: _ * 180. / numpy.pi
```

```
[21]:
```

120.892974293453

### 1.3 Problem 3

Given the following stress tensor

$$\sigma = \begin{bmatrix} 18 & 0 & 24 \\ 0 & -50 & 0 \\ 24 & 0 & 32 \end{bmatrix}$$

Find:

1. the principle stresses  $\sigma_I, \sigma_{II}, \sigma_{III}$
2. the three invariants  $I_1, I_2, I_3$
3. the deviatoric stress
4. the two nonzero invariants of the deviatoric stress, i.e.  $J_2, J_3$

## Solution

Defining the stress tensor

```
[22]: sigma = Matrix([[18, 0, 24],[0, -50, 0],[24, 0, 32]])
```

Here we use sympy to diagonalize (or find the eigenvalues, they are shown on the diagonal of the matrix. We then define  $\sigma_I > \sigma_{II} > \sigma_{III}$  accordingly.

```
[23]: _, D = sigma.diagonalize();  
sigma1 = D[2,2];sigma2 = D[1,1]; sigma3 = D[0,0]; D
```

[23]:

$$\begin{bmatrix} -50 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 50 \end{bmatrix}$$

The first, second, and third invariants

```
[24]: I1 = sigma1 + sigma2 + sigma3; I1
```

[24]:

$$0$$

```
[25]: I2 = sigma1 * sigma2 + sigma1 * sigma3 + sigma2 * sigma3; I2
```

[25]:

$$-2500$$

```
[26]: I3 = sigma1 * sigma2 * sigma3; I3
```

[26]:

$$0$$

The deviatoric stress

```
[27]: Sij = sigma - 1. / 3. * I1 * eye(3); Sij
```

[27]:

$$\begin{bmatrix} 18 & 0 & 24 \\ 0 & -50 & 0 \\ 24 & 0 & 32 \end{bmatrix}$$

Here we perform the same procedure on the deviatoric stress and compute the invariants.

```
[28]: _, D = Sij.diagonalize();  
Sij1 = D[2,2]; Sij2 = D[1,1]; Sij3 = D[0,0];  
  
J2 = Sij1 * Sij2 + Sij1 * Sij3 + Sij2 * Sij3; J2
```

[28]:

$$-2500$$



[29] : J3 = Sij1 \* Sij2 \* Sij3; J3

[29] :

0

#### 1.4 Problem 4

Show that

$$\frac{\partial J_2}{\partial \sigma_{ij}} = S_{ij}$$

where  $J_2$  is the second invariant of the deviatoric stress tensor,  $S_{ij}$ .

**Solution**

$$\frac{\partial J_2}{\partial \sigma_{ij}} = \frac{\partial}{\partial \sigma_{ij}} (J_2) \quad (1)$$

$$= \frac{\partial}{\partial \sigma_{ij}} \left( \frac{1}{2} S_{kl} S_{kl} \right) \quad (2)$$

$$= \frac{1}{2} \left( \frac{\partial}{\partial \sigma_{ij}} (S_{kl}) S_{kl} + S_{kl} \frac{\partial}{\partial \sigma_{ij}} (S_{kl}) \right) \quad (3)$$

$$= S_{kl} \frac{\partial}{\partial \sigma_{ij}} (S_{kl}) \quad (4)$$

$$= S_{kl} \frac{\partial}{\partial \sigma_{ij}} \left( \sigma_{kl} - \frac{1}{3} \sigma_{mm} \delta_{kl} \right) \quad (5)$$

$$= S_{kl} \left( \delta_{il} \delta_{kj} - \frac{1}{3} \delta_{im} \delta_{jm} \delta_{kl} \right) \quad (6)$$

$$= S_{kl} \left( \delta_{il} \delta_{kj} - \frac{1}{3} \delta_{ij} \delta_{kl} \right) \quad (7)$$

$$= S_{ij} - \frac{1}{3} S_{kk} \delta_{ij} \quad (8)$$

$$= S_{ij} \quad (9)$$

because

$$S_{kk} = 0$$

by definition of  $S$  being a deviatoric tensor.

#### 1.5 Problem 5

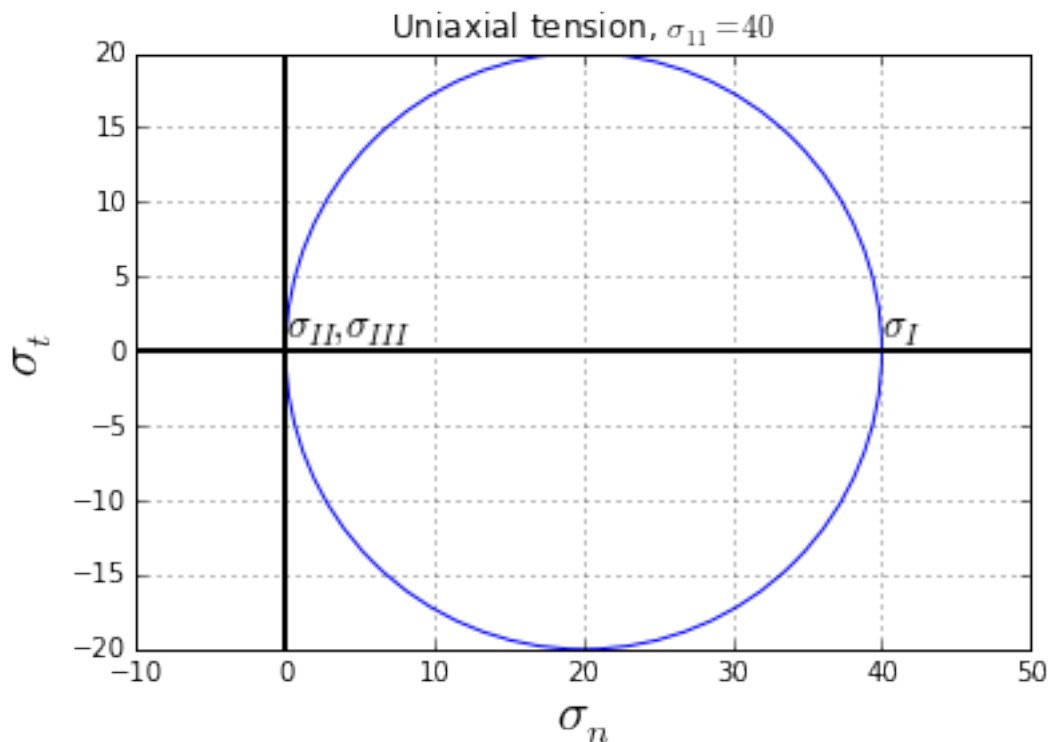
For each of the following stress states (values not given are zero), plot the three Mohr's circles and determine the maximum shear stress.

1. Uniaxial tension  $\sigma_{11} = 40$
2. Biaxial stress  $\sigma_{11} = -10, \sigma_{22} = 30$

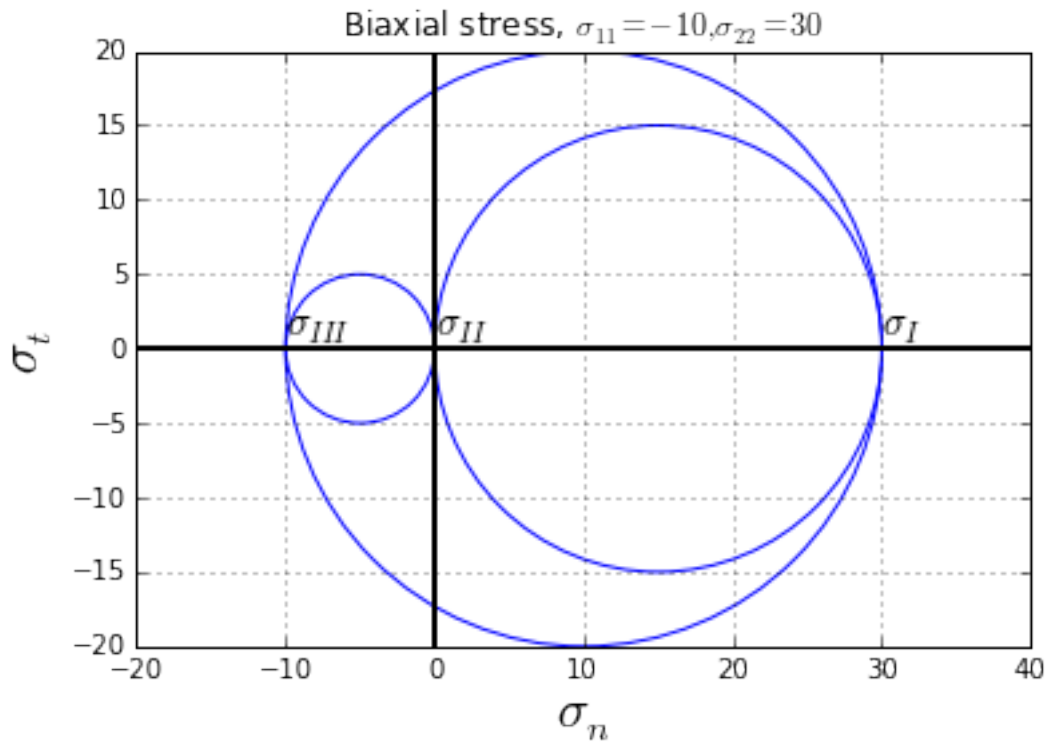
3. Hydrostatic tension of magnitude 100 psi
4.  $\sigma_{11} = -60, \sigma_{22} = 100, \sigma_{33} = 40$
5.  $\sigma_{11} = 10, \sigma_{22} = 40, \sigma_{21} = \sigma_{12} = 20$

### Solution

```
[30]: fig = plt.figure()
ax = fig.add_subplot(111, xlabel='$\sigma_n$',
    ylabel='$\sigma_t$', title='Uniaxial tension, $\sigma_{11}=40$')
ax.xaxis.label.set_size(20)
ax.yaxis.label.set_size(20)
ax.axhline(0, color='black', lw=2)
ax.axvline(0, color='black', lw=2)
ax.annotate('$\sigma_{II}, \sigma_{III}$', xy=(0.04, 1.0), xycoords='data',
    fontsize='15');
ax.annotate('$\sigma_I$', xy=(40, 1.0), xycoords='data', fontsize='15');
circ = plt.Circle((0.5*(40-0), 0), radius=20, fill=False, color='b')
ax.add_patch(circ)
plt.axis('equal')
plt.grid()
plt.show()
```

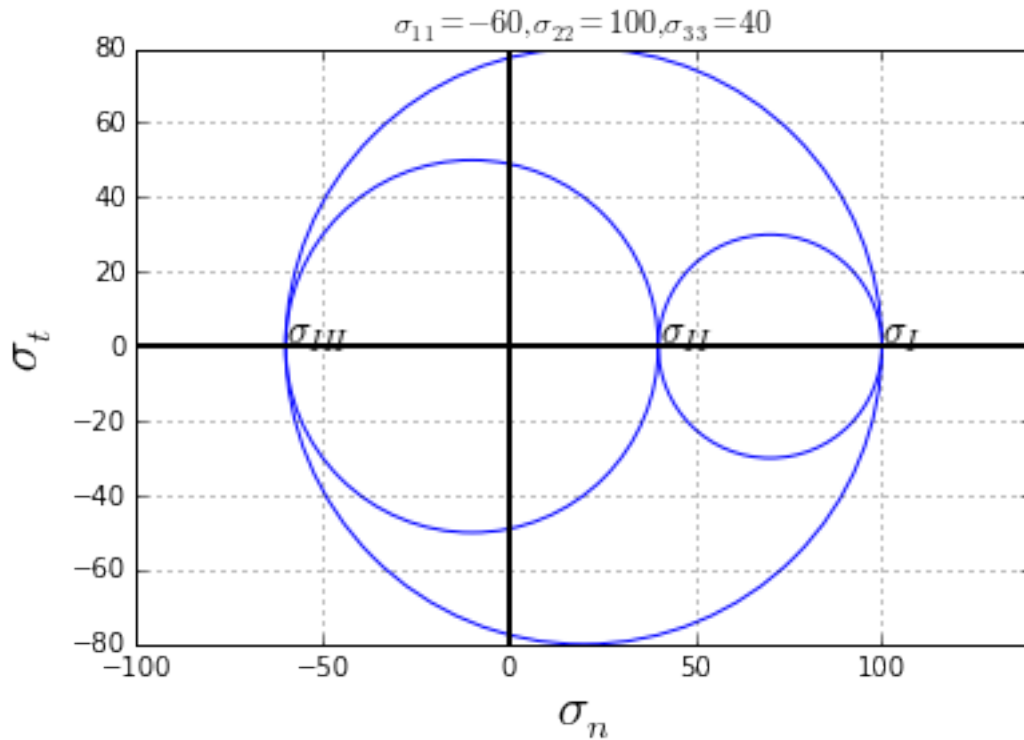


```
[31]: fig = plt.figure()
ax = fig.add_subplot(111,xlabel='$\sigma_n$',
    ylabel='$\sigma_t$',title='Biaxial stress, $\sigma_{11}=-10, \sigma_{22}=30$')
ax.xaxis.label.set_size(20)
ax.yaxis.label.set_size(20)
ax.axhline(0, color='black', lw=2)
ax.axvline(0, color='black', lw=2)
ax.annotate('$\sigma_{III}$', xy=(-10, 1.0), xycoords='data', fontsize='15');
ax.annotate('$\sigma_{II}$', xy=(0.04, 1.0), xycoords='data', fontsize='15');
ax.annotate('$\sigma_I$', xy=(30, 1.0), xycoords='data', fontsize='15');
circ1 = plt.Circle((0.5*(30+(-10))), 0), radius=20, fill=False, color='b')
circ2 = plt.Circle((0.5*(30+(0))), 0), radius=15, fill=False, color='b')
circ3 = plt.Circle((-5, 0), radius=5, fill=False, color='b')
ax.add_patch(circ1)
ax.add_patch(circ2)
ax.add_patch(circ3)
plt.axis('equal')
plt.grid()
plt.show()
```



For the hydrostatic tension case, there are no Mohr's circles to draw because  $\sigma_I = \sigma_{II} = \sigma_{III}$

```
[32]: fig = plt.figure()
ax = fig.add_subplot(111,xlabel='$\sigma_n$', ylabel='$\sigma_t$',
                    title='$\sigma_{11} = -60, \sigma_{22} = 100, \sigma_{33} = 40$')
ax.xaxis.label.set_size(20)
ax.yaxis.label.set_size(20)
ax.axhline(0, color='black', lw=2)
ax.axvline(0, color='black', lw=2)
ax.annotate('$\sigma_{III}$', xy=(-60, 1.0), xycoords='data', fontsize='15');
ax.annotate('$\sigma_{II}$', xy=(40, 1.0), xycoords='data', fontsize='15');
ax.annotate('$\sigma_I$', xy=(100, 1.0), xycoords='data', fontsize='15');
circ1 = plt.Circle((0.5*(100+(-60)), 0), radius=80, fill=False, color='b')
circ2 = plt.Circle((70, 0), radius=30, fill=False, color='b')
circ3 = plt.Circle((-10, 0), radius=50, fill=False, color='b')
ax.add_patch(circ1)
ax.add_patch(circ2)
ax.add_patch(circ3)
plt.axis('equal')
plt.grid()
plt.show()
```



```
[33]: sigma = Matrix([[10,20,0],[20,40,0],[0,0,0]])
_, D = sigma.diagonalize(); D
```

[33]:

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 50 \end{bmatrix}$$

```
[34]: fig = plt.figure()
ax = fig.add_subplot(111,xlabel='$\sigma_n$', ylabel='$\sigma_t$',
                    title='$\sigma_{11} = 10, \sigma_{22} = 40, \sigma_{21} = \sigma_{12} = 20$')
ax.xaxis.label.set_size(20)
ax.yaxis.label.set_size(20)
ax.axhline(0, color='black', lw=2)
ax.axvline(0, color='black', lw=2)
ax.annotate('$\sigma_{II}, \sigma_{III}$', xy=(0.04, 1.0), xycoords='data',
           fontsize='15');
ax.annotate('$\sigma_I$', xy=(50, 1.0), xycoords='data', fontsize='15');
circ = plt.Circle((0.5*(50-0), 0), radius=25, fill=False, color='b')
ax.add_patch(circ)
plt.axis('equal')
plt.grid()
plt.show()
```

