

# Homework 1 Solutions

February 8, 2017

## 1 Problem 1

Composites material properties are usually calibrated in the material coordinate system, with the base vectors denoted as  $a_i$ . To carry out analysis, we usually have to set up problem coordinate system denoted as  $e_i$ . In the most general case, the material coordinate system can be obtained through three consecutive rotations from the problem coordinate system. Let us assume we rotate  $e_i$  about  $e_1$  by  $\theta_1$  to become  $e'_i$ . This is followed by a rotation of  $e'_i$  about  $e'_2$  by  $\theta_2$  to become  $e''_i$ . Finally, we rotate  $e''_i$  about  $e''_3$  by  $\theta_3$  to give  $a_i$ . Find the direction cosine matrix  $\beta_{ij}$  such that  $e_i = \beta_{ij}a_j$ .

### 1.1 Solution

With the convention that

$$Q_{ij} = \cos(x'_i, x_j)$$

We know that

$$e'_i = Q_{ij}e_j$$

We can denote the rotation matrix,  $Q_{ij}$  for each successive transformation with a superscript, giving the following equations

$$e'_i = Q_{ij}e_j \quad e''_i = Q_{ij}^2 e'_j \quad a_i = Q_{ij}^3 e''_j$$

We desire to relate  $a_i$  to  $e_i$  directly, with one effective rotation,  $\beta_{ij}$ . We can do this by substituting the three equations

$$a_i = Q_{ij}^3 Q_{jk}^2 Q_{kl}^1 e_l = [Q^3][Q^2][Q^1]e$$

In this problem, however, we are tasked with finding the inverse relationship,  $e_i = \beta_{ij}a_j$ , left multiply by  $Q^{3T}, Q^{2T}, Q^{1T}$  in that order.

$$Q_{ji}^1 Q_{kj}^2 Q_{lk}^3 a_l = e_i = [Q^{1T}][Q^{2T}][Q^{3T}]$$

to find  $\beta_{ij} = Q_{ji}^1 Q_{kj}^2 Q_{lk}^3$

We can expand this symbolically

```
In [15]: import sympy as sm #python symbolic math library
sm.init_printing() #render output
t1, t2, t3 = sm.symbols(r'\theta_1 \theta_2 \theta_3') #rotation angles
Q1 = sm.Matrix([[1, 0, 0],
                 [0, sm.cos(t1), sm.sin(t1)],
```

```

[0,-sm.sin(t1),sm.cos(t1)])
#Q1*sm.Matrix([0,0,1]) #check rotation signs
Q2 = sm.Matrix([[sm.cos(t2),0,-sm.sin(t2)],
[0,1,0],
[sm.sin(t2),0,sm.cos(t2)])])
#Q2*sm.Matrix([1,0,0]) #check rotation signs
Q3 = sm.Matrix([[sm.cos(t3), sm.sin(t3),0],
[-sm.sin(t3), sm.cos(t3),0],
[0,0,1]])
#Q3*sm.Matrix([0,1,0]) #check rotation signs
beta = Q1.T*Q2.T*Q3.T #.T gives transpose in python
beta

```

Out [15]:

$$\begin{bmatrix} \cos(\theta_2)\cos(\theta_3) & -\sin(\theta_3)\cos(\theta_2) & \sin(\theta_2) \\ \sin(\theta_1)\sin(\theta_2)\cos(\theta_3) + \sin(\theta_3)\cos(\theta_1) & -\sin(\theta_1)\sin(\theta_2)\sin(\theta_3) + \cos(\theta_1)\cos(\theta_3) & -\sin(\theta_1)\cos(\theta_2) \\ \sin(\theta_1)\sin(\theta_3) - \sin(\theta_2)\cos(\theta_1)\cos(\theta_3) & \sin(\theta_1)\cos(\theta_3) + \sin(\theta_2)\sin(\theta_3)\cos(\theta_1) & \cos(\theta_1)\cos(\theta_2) \end{bmatrix}$$

## 1.2 Problem 2

Consider a unidirectional fiber-reinforced composite as linearly elastic, orthotropic materials with

Property	Value
$E_1$	20 Mpsi
$E_2$	1.4 Mpsi
$E_3$	2.5 Mpsi
$G_{12}$	0.8 Mpsi
$G_{13}$	2.0 Mpsi
$G_{23}$	1.5 Mpsi
$\nu_{12}$	0.2
$\nu_{13}$	0.3
$\nu_{23}$	0.25

Where the 1-direction is in the fiber direction, the 2-direction is normal to the fibers in the plane, and the 3-direction is normal to the plane. Four 45° laminae are used to make a box beam (45° refers to the rotation about the outward normal at each wall). Compute the 6x6 stiffness matrix for each wall of the box beam in the global coordinate system.

## 1.3 Solution

First, we calculate the orthotropic stiffness matrix for uni-directional fibers.

```

In [5]: import numpy as np
E1, E2, E3, G12, G13, G23, v12, v13, v23 = 20e6, 1.4e6, 2.5e6, 0.8e6, 2.0e6, 1.5e6, 0.2, 0.3, 0.25
S = np.array([[1/E1, -v12/E1, -v13/E1, 0, 0, 0],
[-v12/E1, 1/E2, -v23/E2, 0, 0, 0],

```

```

        [-v13/E1,-v23/E2,1/E3,0,0,0],
        [0,0,0,1/G23,0,0],
        [0,0,0,0,1/G13,0],
        [0,0,0,0,0,1/G12]])
C = np.linalg.inv(S)
np.round(C*1e-6,decimals=2) #in Mpsi

Out[5]: array([[ 20.41,   0.54,   1.01,   0. ,   0. ,   0. ],
               [  0.54,   1.59,   0.73,   0. ,   0. ,   0. ],
               [  1.01,   0.73,   2.86,   0. ,   0. ,   0. ],
               [  0. ,   0. ,   0. ,   1.5 ,   0. ,   0. ],
               [  0. ,   0. ,   0. ,   0. ,   2. ,   0. ],
               [  0. ,   0. ,   0. ,   0. ,   0. ,   0.8 ]])

```

Now we need to calculate the rotations for each wall. First I will define some common rotation functions to build  $R_\sigma$  consistently for each wall.

```

In [17]: def qij_x(theta):
        """
        rotation tensor about x-axis by some theta
        input: theta (angle in radians)
        output: qij (3x3 rotation tensor)
        """
        qij = np.array([[1,0,0],
                        [0,np.cos(theta),np.sin(theta)],
                        [0,-np.sin(theta),np.cos(theta)]])
        return qij

def qij_y(theta):
        """
        rotation tensor about y-axis by some theta
        input: theta (angle in radians)
        output: qij (3x3 rotation tensor)
        """
        qij = np.array([[np.cos(theta), 0, -np.sin(theta)],
                        [0,1,0],
                        [np.sin(theta),0,np.cos(theta)]])
        return qij

def qij_z(theta):
        """
        rotation tensor about z-axis by some theta
        input: theta (angle in radians)
        output: qij (3x3 rotation tensor)
        """
        qij = np.array([[np.cos(theta),np.sin(theta),0],
                        [-np.sin(theta),np.cos(theta),0],
                        [0,0,1]])

```

```

    return qij

def R_sigma(q):
    """
    rotation matrix for engineering notation given some rotation tensor, q
    note: uses convention for sigma = [s11, s22, s33, s23, s13, s12]
    input: q (3x3 rotation tensor)
    output: R_s (6x6 rotation matrix)
    """
    R_s = np.array([[q[0,0]**2, q[0,1]**2, q[0,2]**2, 2*q[0,1]*q[0,2], 2*q[0,0]*q[0,2], 2*q[0,1]*q[0,0],
                    [q[1,0]**2, q[1,1]**2, q[1,2]**2, 2*q[1,1]*q[1,2], 2*q[1,0]*q[1,2], 2*q[1,1]*q[1,0],
                    [q[2,0]**2, q[2,1]**2, q[2,2]**2, 2*q[2,1]*q[2,2], 2*q[2,0]*q[2,2], 2*q[2,1]*q[2,0],
                    [q[1,0]*q[2,0], q[1,1]*q[2,1], q[1,2]*q[2,2], q[1,1]*q[2,0], q[1,0]*q[2,1], q[1,2]*q[2,0],
                    [q[0,0]*q[2,0], q[0,1]*q[2,1], q[0,2]*q[2,2], q[0,1]*q[2,0], q[0,0]*q[2,1], q[0,2]*q[2,0],
                    [q[0,0]*q[1,0], q[0,1]*q[1,1], q[0,2]*q[1,2], q[0,1]*q[1,0], q[0,0]*q[1,1], q[0,2]*q[1,0]]])

    return R_s

```

We will start by considering the face that intersects the positive  $x_2$  axis. Since the lamina is orthotropic, there are multiple local coordinate systems that would give the same properties.

Recall that, as defined in Problem 1, we have  $e_i = \beta_{ij}a_j$ , so we can use  $\beta_{ij}$  from Problem 1 to directly relate the local coordinate system (fiber direction with the fibers pointing along  $x'_1$  and the outward normal being  $x'_3$ ) to the global coordinate system.

We can use the  $\beta$  defined in Problem 1 by first rotating about  $x_1$  until  $x_3$  is the normal, then rotating by  $45^\circ$  about the normal ( $x'_3$ ).

Starting at the right wall, we have  $\theta_1 = 270$  (or  $-90$ ) and  $\theta_3 = 45$  (with  $\theta_2 = 0$ ).

We can check this rotation by multiplying  $\beta$  by  $(1,0,0)$ ,  $(0,1,0)$  and  $(0,0,1)$  (the unit vectors in the prime coordinates), this should give back global coordinates of  $(1,0,-1)/\sqrt{2}$ ,  $(-1,0,-1)/\sqrt{2}$ , and  $(0,1,0)$  respectively.

```

In [28]: beta_rightwall = beta.subs([(t1,-sm.pi/2), (t2,0), (t3,sm.pi/4)])
        #beta_rightwall*sm.Matrix([0,0,1])

```

Now we need to be careful about the  $R_\sigma$  as defined. The convention used for this  $R_\sigma$  is that  $R_\sigma$  is a function of  $Q_{ij}$ , and gives the relationship

$$C' = R_\sigma C R_\sigma^T$$

However, we already know  $C'$  and desire to find  $C$  in the global reference. We can achieve this by either re-writing our equation, or calculating  $R_\sigma$  with  $Q^{-1} = Q^T = \beta$

```

In [70]: #convert beta to numeric python library
        B = np.array(beta_rightwall.tolist()).astype(np.float64)
        R = R_sigma(B)
        print np.round(np.dot(R, np.dot(C, R.T)) * 1e-6, decimals=2)

```

```

[[ 6.57  0.87  4.97  0.   -4.7   0. ]
 [ 0.87  2.86  0.87  0.   -0.14  0. ]
 [ 4.97  0.87  6.57  0.   -4.7   0. ]
 [ 0.    0.    0.    1.75  0.   -0.25]

```

```
[-4.7  -0.14 -4.7   0.    5.23  0. ]
[ 0.    0.    0.   -0.25  0.    1.75]]
```

We can now look at the top wall, where  $x_3$  is already normal, so we simply set  $\theta_3 = 45^\circ$  with  $\theta_1 = \theta_2 = 0$

```
In [71]: beta_topwall = beta.subs([(t1,0),(t2,0),(t3,sm.pi/4)])
        #print beta_topwall*sm.Matrix([0,0,1]) #check rotations
        B = np.array(beta_topwall.tolist()).astype(np.float64)
        R = R_sigma(B)
        print np.round(np.dot(R,np.dot(C,R.T))*1e-6,decimals=2)

[[ 6.57  4.97  0.87  0.    0.    4.7 ]
 [ 4.97  6.57  0.87  0.    0.    4.7 ]
 [ 0.87  0.87  2.86  0.    0.    0.14]
 [ 0.    0.    0.    1.75  0.25  0. ]
 [ 0.    0.    0.    0.25  1.75  0. ]
 [ 4.7   4.7   0.14  0.    0.    5.23]]
```

And for the left wall we set  $\theta_1 = 90$  with  $\theta_2 = 0$  and  $\theta_3 = 45$

```
In [72]: beta_leftwall = beta.subs([(t1,sm.pi/2),(t2,0),(t3,sm.pi/4)])
        #beta_leftwall*sm.Matrix([0,1,0]) #check rotations
        B = np.array(beta_leftwall.tolist()).astype(np.float64)
        R = R_sigma(B)
        C_left = np.dot(R,np.dot(C,R.T)) #save for problem 3
        print np.round(C_left*1e-6,decimals=2)

[[ 6.57  0.87  4.97  0.    4.7  0. ]
 [ 0.87  2.86  0.87  0.    0.14  0. ]
 [ 4.97  0.87  6.57  0.    4.7  0. ]
 [ 0.    0.    0.    1.75  0.    0.25]
 [ 4.7   0.14  4.7   0.    5.23  0. ]
 [ 0.    0.    0.    0.25  0.    1.75]]
```

Finally for the bottom wall we have  $\theta_1 = 180, \theta_2 = 0$  and  $\theta_3 = 45$

```
In [73]: beta_bottomwall = beta.subs([(t1,sm.pi),(t2,0),(t3,sm.pi/4)])
        #beta_bottomwall*sm.Matrix([0,1,0]) #check rotations
        B = np.array(beta_bottomwall.tolist()).astype(np.float64)
        R = R_sigma(B)
        print np.round(np.dot(R,np.dot(C,R.T))*1e-6,decimals=2)

[[ 6.57  4.97  0.87  0.    0.   -4.7 ]
 [ 4.97  6.57  0.87  0.    0.   -4.7 ]
 [ 0.87  0.87  2.86  0.    0.   -0.14]
 [ 0.    0.    0.    1.75 -0.25  0. ]
 [ 0.    0.    0.   -0.25  1.75  0. ]
 [-4.7  -4.7  -0.14  0.    0.    5.23]]
```

## 2 Problem 3

To find  $R_\sigma$  and  $R_\epsilon$  we start by multiplying out the transformation relationship.

```
In [74]: #define symbols for stress tensor
s1, s2, s3, s4, s5, s6 = sm.symbols('\sigma_1 \sigma_2 \sigma_3 \sigma_4
s = sm.Matrix([[s1, s6, s5],
               [s6, s2, s4],
               [s5, s4, s3]]) #symmetric stress tensor in engineering notation
Q = sm.Matrix(3, 3, lambda i,j:sm.symbols('Q_{%d%d}' % (i+1,j+1))) #symbolic
sp = Q*s*Q.T
sp = sp.expand() #multiply out terms
#write matrix in vector form in correct order
spe = sm.Matrix([[sp[0,0]],
                 [sp[0,1]],
                 [sp[1,1]],
                 [sp[0,2]],
                 [sp[1,2]],
                 [sp[2,2]]]) #reshape to vector for engineering notation
spe #notice that terms are not necessarily sorted, nor are they combined
```

Out [74]:

$$\begin{bmatrix} Q_{11}^2\sigma_1 + 2Q_{11}Q_{12}\sigma_6 + 2Q_{11}Q_{13}\sigma_5 + Q_{12}^2\sigma_2 + 2Q_{12}Q_{13}\sigma_4 + Q_{13}^2\sigma_3 \\ Q_{11}Q_{21}\sigma_1 + Q_{11}Q_{22}\sigma_6 + Q_{11}Q_{23}\sigma_5 + Q_{12}Q_{21}\sigma_6 + Q_{12}Q_{22}\sigma_2 + Q_{12}Q_{23}\sigma_4 + Q_{13}Q_{21}\sigma_5 + Q_{13}Q_{22}\sigma_4 + Q_{13}Q_{23}\sigma_3 \\ Q_{21}^2\sigma_1 + 2Q_{21}Q_{22}\sigma_6 + 2Q_{21}Q_{23}\sigma_5 + Q_{22}^2\sigma_2 + 2Q_{22}Q_{23}\sigma_4 + Q_{23}^2\sigma_3 \\ Q_{11}Q_{31}\sigma_1 + Q_{11}Q_{32}\sigma_6 + Q_{11}Q_{33}\sigma_5 + Q_{12}Q_{31}\sigma_6 + Q_{12}Q_{32}\sigma_2 + Q_{12}Q_{33}\sigma_4 + Q_{13}Q_{31}\sigma_5 + Q_{13}Q_{32}\sigma_4 + Q_{13}Q_{33}\sigma_3 \\ Q_{21}Q_{31}\sigma_1 + Q_{21}Q_{32}\sigma_6 + Q_{21}Q_{33}\sigma_5 + Q_{22}Q_{31}\sigma_6 + Q_{22}Q_{32}\sigma_2 + Q_{22}Q_{33}\sigma_4 + Q_{23}Q_{31}\sigma_5 + Q_{23}Q_{32}\sigma_4 + Q_{23}Q_{33}\sigma_3 \\ Q_{31}^2\sigma_1 + 2Q_{31}Q_{32}\sigma_6 + 2Q_{31}Q_{33}\sigma_5 + Q_{32}^2\sigma_2 + 2Q_{32}Q_{33}\sigma_4 + Q_{33}^2\sigma_3 \end{bmatrix}$$

Now we need to write this as a matrix equation by factoring out like terms.

```
In [75]: #here we will sort and combine terms
R_s = sm.zeros(6) #first fill with zeros
cols = [s1,s6,s2,s5,s4,s3] #define term order
#loop through all values of R_s
for i in range(6):
    row = spe[i] #find row of matrix to work with
    for j in range(6):
        col = cols[j]
        temp = sm.collect(row,col,evaluate=False) #collect terms in row with
        R_s[i,j] = temp[col] #factor out term from col, and insert it into
R_s
```

Out [75]:

$$\begin{bmatrix} Q_{11}^2 & 2Q_{11}Q_{12} & Q_{12}^2 & 2Q_{11}Q_{13} & 2Q_{12}Q_{13} & Q_{13}^2 \\ Q_{11}Q_{21} & Q_{11}Q_{22} + Q_{12}Q_{21} & Q_{12}Q_{22} & Q_{11}Q_{23} + Q_{13}Q_{21} & Q_{12}Q_{23} + Q_{13}Q_{22} & Q_{13}Q_{23} \\ Q_{21}^2 & 2Q_{21}Q_{22} & Q_{22}^2 & 2Q_{21}Q_{23} & 2Q_{22}Q_{23} & Q_{23}^2 \\ Q_{11}Q_{31} & Q_{11}Q_{32} + Q_{12}Q_{31} & Q_{12}Q_{32} & Q_{11}Q_{33} + Q_{13}Q_{31} & Q_{12}Q_{33} + Q_{13}Q_{32} & Q_{13}Q_{33} \\ Q_{21}Q_{31} & Q_{21}Q_{32} + Q_{22}Q_{31} & Q_{22}Q_{32} & Q_{21}Q_{33} + Q_{23}Q_{31} & Q_{22}Q_{33} + Q_{23}Q_{32} & Q_{23}Q_{33} \\ Q_{31}^2 & 2Q_{31}Q_{32} & Q_{32}^2 & 2Q_{31}Q_{33} & 2Q_{32}Q_{33} & Q_{33}^2 \end{bmatrix}$$

We can find the transformation for engineering strain by expanding the transformation for strain, but replacing the shear strain terms  $\epsilon_{12}$  with engineering strain divided by 2  $\gamma_{12}/2$ . We can then multiply both sides of the shear strain equations by two to find the transformation in terms of the engineering strain.

```
In [76]: #define strain tensor symbols
e1, e2, e3, e4, e5, e6 = sm.symbols('\epsilon_1 \epsilon_2 \epsilon_3 \epsilon_4 \epsilon_5 \epsilon_6')
e = sm.Matrix([[e1, e6, e5],
               [e6, e2, e4],
               [e5, e4, e3]])
#engineering strain symbols
g12, g13, g23 = sm.symbols('\gamma_{12} \gamma_{13} \gamma_{23}')
#substitute engineering strain
e=e.subs([(e4,g23/2),(e5,g13/2),(e6,g12/2)])
e
```

Out [76]:

$$\begin{bmatrix} \epsilon_1 & \frac{\gamma_{12}}{2} & \frac{\gamma_{13}}{2} \\ \frac{\gamma_{12}}{2} & \epsilon_2 & \frac{\gamma_{23}}{2} \\ \frac{\gamma_{13}}{2} & \frac{\gamma_{23}}{2} & \epsilon_3 \end{bmatrix}$$

```
In [77]: #transformation equation
ep = Q*e*Q.T
ep = ep.expand()
ep
```

Out [77]:

$$\begin{bmatrix} Q_{11}^2\epsilon_1 + Q_{11}Q_{12}\gamma_{12} + Q_{11}Q_{13}\gamma_{13} + Q_{12}^2\epsilon_2 + Q_{12}Q_{13}\gamma_{23} + Q_{13}^2\epsilon_3 & 2Q_{11}Q_{21}\epsilon_1 + Q_{11}Q_{22}\gamma_{12} + Q_{11}Q_{23}\gamma_{13} + Q_{12}Q_{22}\epsilon_2 + Q_{12}Q_{23}\gamma_{23} + Q_{13}Q_{21}\gamma_{13} + Q_{13}Q_{22}\gamma_{23} + Q_{13}Q_{23}\epsilon_3 \\ 2Q_{11}Q_{31}\epsilon_1 + Q_{11}Q_{32}\gamma_{12} + Q_{11}Q_{33}\gamma_{13} + Q_{12}Q_{32}\epsilon_2 + Q_{12}Q_{33}\gamma_{23} + Q_{13}Q_{31}\gamma_{13} + Q_{13}Q_{32}\gamma_{23} + Q_{13}Q_{33}\epsilon_3 \end{bmatrix}$$

Notice that the shear equations show the transformation for  $\epsilon'_{12}$ , or  $\gamma'_{12}/2$ . Since we want a transformation for  $\gamma'_{12}$ , we will multiply both sides of those equations by 2.

```
In [78]: #write matrix in vector form in correct order
epe = sm.Matrix([[ep[0,0]],
                 [2*ep[0,1]],
                 [ep[1,1]],
```

```

[2*ep[0,2]],
[2*ep[1,2]],
[ep[2,2]]) #reshape to vector for engineering notation
epe #notice that terms are not necessarily sorted, nor are they combined

```

Out [78]:

$$\begin{bmatrix}
Q_{11}^2 \epsilon_1 + Q_{11} Q_{12} \gamma_{12} + Q_{11} Q_{13} \gamma_{13} + Q_{12}^2 \epsilon_2 + Q_{12} Q_{13} \gamma_{23} + Q_{13}^2 \epsilon_3 \\
2Q_{11} Q_{21} \epsilon_1 + Q_{11} Q_{22} \gamma_{12} + Q_{11} Q_{23} \gamma_{13} + Q_{12} Q_{21} \gamma_{12} + 2Q_{12} Q_{22} \epsilon_2 + Q_{12} Q_{23} \gamma_{23} + Q_{13} Q_{21} \gamma_{13} + Q_{13} Q_{22} \gamma_{23} + 2Q_{13} Q_{23} \epsilon_3 \\
Q_{21}^2 \epsilon_1 + Q_{21} Q_{22} \gamma_{12} + Q_{21} Q_{23} \gamma_{13} + Q_{22}^2 \epsilon_2 + Q_{22} Q_{23} \gamma_{23} + Q_{23}^2 \epsilon_3 \\
2Q_{11} Q_{31} \epsilon_1 + Q_{11} Q_{32} \gamma_{12} + Q_{11} Q_{33} \gamma_{13} + Q_{12} Q_{31} \gamma_{12} + 2Q_{12} Q_{32} \epsilon_2 + Q_{12} Q_{33} \gamma_{23} + Q_{13} Q_{31} \gamma_{13} + Q_{13} Q_{32} \gamma_{23} + 2Q_{13} Q_{33} \epsilon_3 \\
2Q_{21} Q_{31} \epsilon_1 + Q_{21} Q_{32} \gamma_{12} + Q_{21} Q_{33} \gamma_{13} + Q_{22} Q_{31} \gamma_{12} + 2Q_{22} Q_{32} \epsilon_2 + Q_{22} Q_{33} \gamma_{23} + Q_{23} Q_{31} \gamma_{13} + Q_{23} Q_{32} \gamma_{23} + 2Q_{23} Q_{33} \epsilon_3 \\
Q_{31}^2 \epsilon_1 + Q_{31} Q_{32} \gamma_{12} + Q_{31} Q_{33} \gamma_{13} + Q_{32}^2 \epsilon_2 + Q_{32} Q_{33} \gamma_{23} + Q_{33}^2 \epsilon_3
\end{bmatrix}$$

Now we can sort terms and factor into a matrix equation

```

In [79]: R_e = sm.zeros(6) #first fill with zeros
cols = [e1,g12,e2,g13,g23,e3] #define term order
#loop through all values of R_s
for i in range(6):
    row = epe[i] #find row of matrix to work with
    for j in range(6):
        col = cols[j]
        temp = sm.collect(row,col,evaluate=False) #collect terms in row with
        R_e[i,j] = temp[col] #factor out term from col, and insert it into
R_e

```

Out [79]:

$$\begin{bmatrix}
Q_{11}^2 & Q_{11} Q_{12} & Q_{12}^2 & Q_{11} Q_{13} & Q_{12} Q_{13} & Q_{13}^2 \\
2Q_{11} Q_{21} & Q_{11} Q_{22} + Q_{12} Q_{21} & 2Q_{12} Q_{22} & Q_{11} Q_{23} + Q_{13} Q_{21} & Q_{12} Q_{23} + Q_{13} Q_{22} & 2Q_{13} Q_{23} \\
Q_{21}^2 & Q_{21} Q_{22} & Q_{22}^2 & Q_{21} Q_{23} & Q_{22} Q_{23} & Q_{23}^2 \\
2Q_{11} Q_{31} & Q_{11} Q_{32} + Q_{12} Q_{31} & 2Q_{12} Q_{32} & Q_{11} Q_{33} + Q_{13} Q_{31} & Q_{12} Q_{33} + Q_{13} Q_{32} & 2Q_{13} Q_{33} \\
2Q_{21} Q_{31} & Q_{21} Q_{32} + Q_{22} Q_{31} & 2Q_{22} Q_{32} & Q_{21} Q_{33} + Q_{23} Q_{31} & Q_{22} Q_{33} + Q_{23} Q_{32} & 2Q_{23} Q_{33} \\
Q_{31}^2 & Q_{31} Q_{32} & Q_{32}^2 & Q_{31} Q_{33} & Q_{32} Q_{33} & Q_{33}^2
\end{bmatrix}$$

To verify that  $(R_\epsilon)^{-1} = R_\sigma^T$  for  $\theta_1 = 20$ ,  $\theta_2 = 45$  and  $\theta_3 = 60$  we can substitute into the beta we found previously, calculate  $Q = \beta^T$ , then substitute values for both  $R_\epsilon$  and  $R_\sigma$ .

```

In [80]: b3 = beta.subs([(t1, sm.pi/9.), (t2, sm.pi/4.), (t3, sm.pi/3)])
q = np.array(b3.tolist()).astype(np.float64)
q = q.T
#add new matrix to store values for numeric R_e and R_s
R_s_vals = R_s
R_e_vals = R_e
#substitute into R_s and R_e
for i in range(3):
    for j in range(3):

```



```

        R_e_vals = R_e_vals.subs(Q[i,j],q[i,j])
        R_s_vals = R_s_vals.subs(Q[i,j],q[i,j])
R_s_vals = np.array(R_s_vals.tolist()).astype(np.float64)
R_e_vals = np.array(R_e_vals.tolist()).astype(np.float64)
np.round(np.linalg.inv(R_e_vals)-R_s_vals.T,decimals=2)

```

```

Out[80]: array([[ 0.,  0.,  0.,  0.,  0.,  0.],
 [ 0.,  0.,  0.,  0.,  0.,  0.],
 [ 0.,  0.,  0.,  0.,  0.,  0.],
 [ 0.,  0.,  0.,  0.,  0.,  0.],
 [ 0.,  0.,  0.,  0.,  0.,  0.],
 [ 0.,  0.,  0.,  0.,  0.,  0.]])

```

To transform the stiffness in this new notation, we first need to build the uni-directional stiffness matrix in this notation

```

In [81]: E1, E2, E3, G12, G13, G23, v12, v13, v23 = 20e6, 1.4e6, 2.5e6, 0.8e6, 2.0e6, 0.3e6, 0.4e6, 0.5e6
S_new = np.array([[1/E1, 0, -v12/E1, 0, 0, -v13/E1],
 [0, 1/G12, 0, 0, 0, 0],
 [-v12/E1, 0, 1/E2, 0, 0, -v23/E2],
 [0, 0, 0, 1/G13, 0, 0],
 [0, 0, 0, 0, 1/G23, 0],
 [-v13/E1, 0, -v23/E2, 0, 0, 1/E3]])
C_new = np.linalg.inv(S_new)
np.round(C_new*1e-6,decimals=2) #in Mpsi

```

```

Out[81]: array([[ 20.41,  0. ,  0.54,  0. ,  0. ,  1.01],
 [  0. ,  0.8 ,  0. ,  0. ,  0. ,  0. ],
 [  0.54,  0. ,  1.59,  0. ,  0. ,  0.73],
 [  0. ,  0. ,  0. ,  2. ,  0. ,  0. ],
 [  0. ,  0. ,  0. ,  0. ,  1.5 ,  0. ],
 [  1.01,  0. ,  0.73,  0. ,  0. ,  2.86]])

```

Now we perform the rotation using the  $\beta$  and  $Q$  we assembled for the left wall

```

In [82]: b3 = beta.subs([(t1,sm.pi/2),(t2,0),(t3,sm.pi/4)])
q = np.array(b3.tolist()).astype(np.float64)
#add new matrix to store values for numeric R_e and R_s
R_s_vals = R_s
R_e_vals = R_e
#substitute into R_s and R_e
for i in range(3):
    for j in range(3):
        R_s_vals = R_s_vals.subs(Q[i,j],q[i,j])
R_s_vals = np.array(R_s_vals.tolist()).astype(np.float64)

C_left_new = np.dot(R_s_vals,np.dot(C_new,R_s_vals.T))
print np.round(C_left_new*1e-6,decimals=2)

```

```
[[ 6.57  0.    0.87  4.7   0.    4.97]
 [ 0.    1.75  0.    0.    0.25  0.   ]
 [ 0.87  0.    2.86  0.14  0.    0.87]
 [ 4.7   0.    0.14  5.23  0.    4.7  ]
 [ 0.    0.25  0.    0.    1.75  0.   ]
 [ 4.97  0.    0.87  4.7   0.    6.57]]
```

Now we can confirm that  $C' = BCB^T$

```
In [83]: B = np.array([[1,0,0,0,0,0],
                        [0,0,0,0,0,1],
                        [0,1,0,0,0,0],
                        [0,0,0,0,1,0],
                        [0,0,0,1,0,0],
                        [0,0,1,0,0,0]])
np.round(np.dot(B,np.dot(C_left,B.T))*1e-6,decimals=2)

Out[83]: array([[ 6.57,  0.   ,  0.87,  4.7  ,  0.   ,  4.97],
 [ 0.   ,  1.75,  0.   ,  0.   ,  0.25,  0.   ],
 [ 0.87,  0.   ,  2.86,  0.14,  0.   ,  0.87],
 [ 4.7  ,  0.   ,  0.14,  5.23,  0.   ,  4.7  ],
 [ 0.   ,  0.25,  0.   ,  0.   ,  1.75,  0.   ],
 [ 4.97,  0.   ,  0.87,  4.7  ,  0.   ,  6.57]])
```