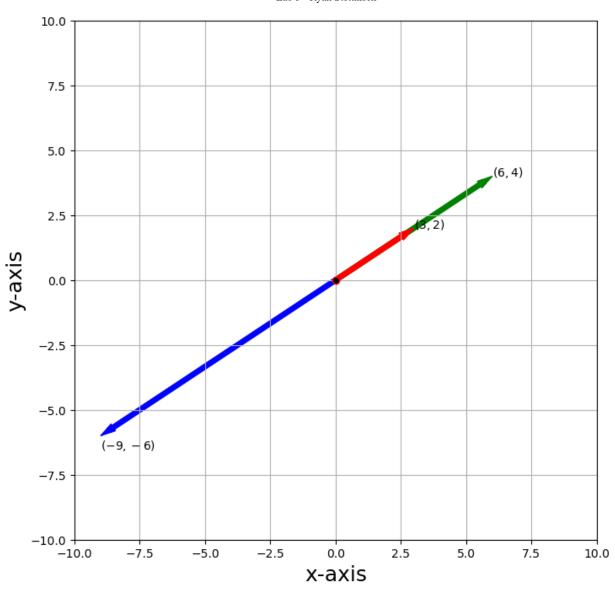
```
In [1]: #1
        print('Python Lab 5 - Ryan Stettnisch')
        import numpy as np
        Python Lab 5 - Ryan Stettnisch
In [2]:
        import numpy as np
        def gram schmidt(a):
            q = []
            for i in range(len(a)):
                #orthogonalization
                q_tilde = a[i]
                for j in range(len(q)):
                    q \text{ tilde} = q \text{ tilde} - (q[j] @ a[i])*q[j]
                #Test for dependennce
                if np.sqrt(sum(q_tilde**2)) <= 1e-10:</pre>
                    print('Vectors are linearly dependent.')
                    print('GS algorithm terminates at iteration ', i+1)
                    return q
                #Normalization
                else:
                    q tilde = q tilde / np.sqrt(sum(q tilde**2))
                    q.append(q_tilde)
            print('Vectors are linearly independent.')
            return q
In [3]: #2a
        a = np.array([[1,-1.1],[-2.8,-0.3],[-0.4, 1.5]]) #This is linearly dependent be
        print(gram schmidt(a))
        Vectors are linearly dependent.
        GS algorithm terminates at iteration 3
        [array([ 0.67267279, -0.73994007]), array([-0.73994007, -0.67267279])]
In [4]: #2b
        b = np.array([[1,0,1,0,1],[0,1,0,0,1],[0,1,0,1,0]])
        print(gram schmidt(b))
        Vectors are linearly independent.
                                    , 0.57735027, 0.
        [array([0.57735027, 0.
                                                              , 0.57735027]), array([-
        0.25819889, 0.77459667, -0.25819889, 0. , 0.51639778]), array([ 0.16
        903085, 0.3380617, 0.16903085, 0.84515425, -0.3380617])]
In [5]: #2c
        c = np.array([[1,2,0],[-2,0,3],[1,0,2]])
        print(gram schmidt(c))
        Vectors are linearly independent.
        [array([0.4472136 , 0.89442719, 0.
                                             ]), array([-0.45807867, 0.22903933,
        0.8588975 ]), array([ 0.76822128, -0.38411064, 0.51214752])]
In [6]: #2d
        d = np.array([[1,2,0],[-2,0,3],[-1,2,3]])
        print(gram schmidt(d))
```

```
Vectors are linearly dependent.
         GS algorithm terminates at iteration 3
         [array([0.4472136 , 0.89442719, 0.
                                                  ]), array([-0.45807867, 0.22903933,
         0.8588975 ])]
In [7]: #3a
         q = gram_schmidt(a)
         print('Norm of q[0] : ',(sum(q[0]**2))**0.5)
         print('Inner product of q[0] and q[1] : ', q[0] @ q[1])
         print('Norm of q[1] : ', (sum(q[1]**2))**0.5)
         Vectors are linearly dependent.
         GS algorithm terminates at iteration 3
         Norm of q[0]: 1.0
         Inner product of q[0] and q[1]: 0.0
         Norm of q[1]: 1.0
 In [8]: #3b
         q = gram_schmidt(b)
         print('Norm of q[0] :', (sum(q[0]**2))**0.5)
         print('Inner product of q[0] and q[1] :', q[0] @ q[1])
         print('Inner product of q[0] and q[2] :', q[0] @ q[2])
         print('Norm of q[1] :', (sum(q[1]**2))**0.5)
         print('Inner product of q[1] and q[2] :', q[1] @ q[2])
         print('Norm of q[2] :', (sum(q[2]**2))**0.5)
         Vectors are linearly independent.
         Norm of q[0] : 1.0
         Inner product of q[0] and q[1] : -1.1102230246251565e-16
         Inner product of q[0] and q[2]: 8.326672684688674e-17
         Norm of q[1] : 1.0
         Inner product of q[1] and q[2]: -1.1102230246251565e-16
         Norm of q[2] : 1.0
 In [9]: #3c
         q = gram schmidt(c)
         print('Norm of q[0] :', (sum(q[0]**2))**0.5)
         print('Inner product of q[0] and q[1] :', q[0] @ q[1])
         print('Inner product of q[0] and q[2] :', q[0] @ q[2])
         print('Norm of q[1] :', (sum(q[1]**2))**0.5)
         print('Inner product of q[1] and q[2] :', q[1] @ q[2])
         print('Norm of q[2] :', (sum(q[2]**2))**0.5)
         Vectors are linearly independent.
         Norm of q[0] : 1.0
         Inner product of q[0] and q[1]: -2.7755575615628914e-17
         Inner product of q[0] and q[2]: 5.551115123125783e-17
         Norm of q[1] : 1.0
         Inner product of q[1] and q[2]: 0.0
         Norm of q[2] : 1.0
In [10]: #3d
         q = gram_schmidt(d)
         print('Norm of q[0] :', (sum(q[0]**2))**0.5)
         print('Inner product of q[0] and q[1] :', q[0] @ q[1])
         print('Norm of q[1] :', (sum(q[1]**2))**0.5)
```

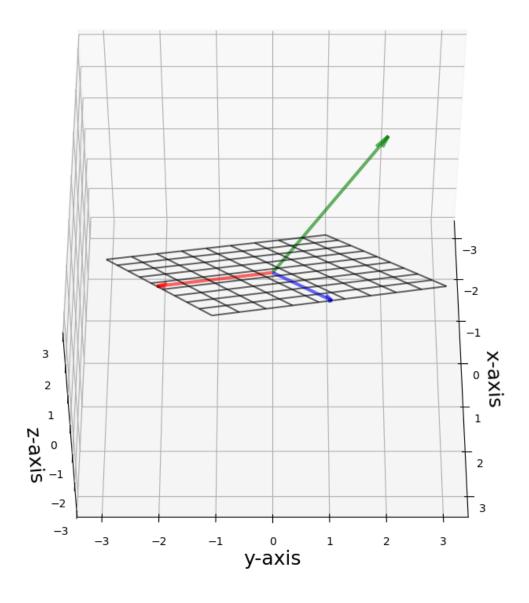
```
Vectors are linearly dependent. GS algorithm terminates at iteration 3 Norm of q[0]: 1.0 Inner product of q[0] and q[1]: -2.7755575615628914e-17 Norm of q[1]: 1.0
```

```
In [11]:
         #4
         import matplotlib.pyplot as plt
         from mpl_toolkits.mplot3d import Axes3D
         import sympy as sy
         sy.init printing()
         fig, ax = plt.subplots(figsize = (8, 8))
         arrows = np.array([[[0,0,3,2]],
         [[0,0,-9,-6]],
         [[0,0,6,4]]])
         colors = ['r','b','g']
         for i in range(arrows.shape[0]):
            X,Y,U,V = zip(*arrows[i,:,:])
             ax.arrow(X[0], Y[0], U[0], V[0], color = colors[i], width = .18,
                 length_includes_head = True,
                head_width = .3, # default: 3*width
                head_length = .6,
                overhang = .4, zorder = -i)
         ax.scatter(0, 0, ec = "red", fc = "black", zorder = 5)
         ax.text(6, 4, '$(6, 4)$')
         ax.text(-9, -6.5, '$(-9, -6)$')
         ax.text(3, 2, '$(3, 2)$')
         ax.grid(True)
         ax.axis([-10, 10, -10, 10])
         ax.set xlabel('x-axis', size = 18)
         ax.set ylabel('y-axis', size = 18)
         plt.show()
         #These vectors are linearly independent
         #You can reach any vector that is a multiple of (3,2) like (6,4), (9,6)
```



```
In [12]:
         #5
         fig = plt.figure(figsize = (10,10))
         ax = fig.add subplot(projection='3d')
         s = np.linspace(-1, 1, 10)
         t = np.linspace(-1, 1, 10)
         S, T = np.meshgrid(s, t)
         X = S+2*T
         Y = -2*S+T
         Z = S+2*T
         ax.plot_wireframe(X, Y, Z, linewidth = 1.5, color = 'k', alpha = .6)
         vec = np.array([[[0, 0, 0, 1, -2, 1]],
                         [[0, 0, 0, 2, 1, 2]],
                         [[0, 0, 0, -1, 2, 3]])
         colors = ['r','b','g']
         for i in range(vec.shape[0]):
             X, Y, Z, U, V, W = zip(*vec[i,:,:])
             ax.quiver(X, Y, Z, U, V, W, length=1, normalize=False, color = colors[i],
                 arrow_length_ratio = .08, pivot = 'tail',
                 linestyles = 'solid',linewidths = 3, alpha = .6)
         ax.set xlabel('x-axis', size = 18)
```

```
ax.set_ylabel('y-axis', size = 18)
ax.set_zlabel('z-axis', size = 18)
ax.view_init(elev=50., azim=0)
plt.show()
#Linearly independent since you can reach any vector where
# x = 1a + 2b - 1c
# y = -2a + 1b + 2c
# z = 1a + 2b + 3c
```



In [ ]: