

Principal Component Analysis

import library

In [1]:



```
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.colors as colors
from matplotlib import cm
```

load data

In [2]:



```
fname_data = 'assignment_12_data.txt'

feature = np.genfromtxt(fname_data, delimiter=',')

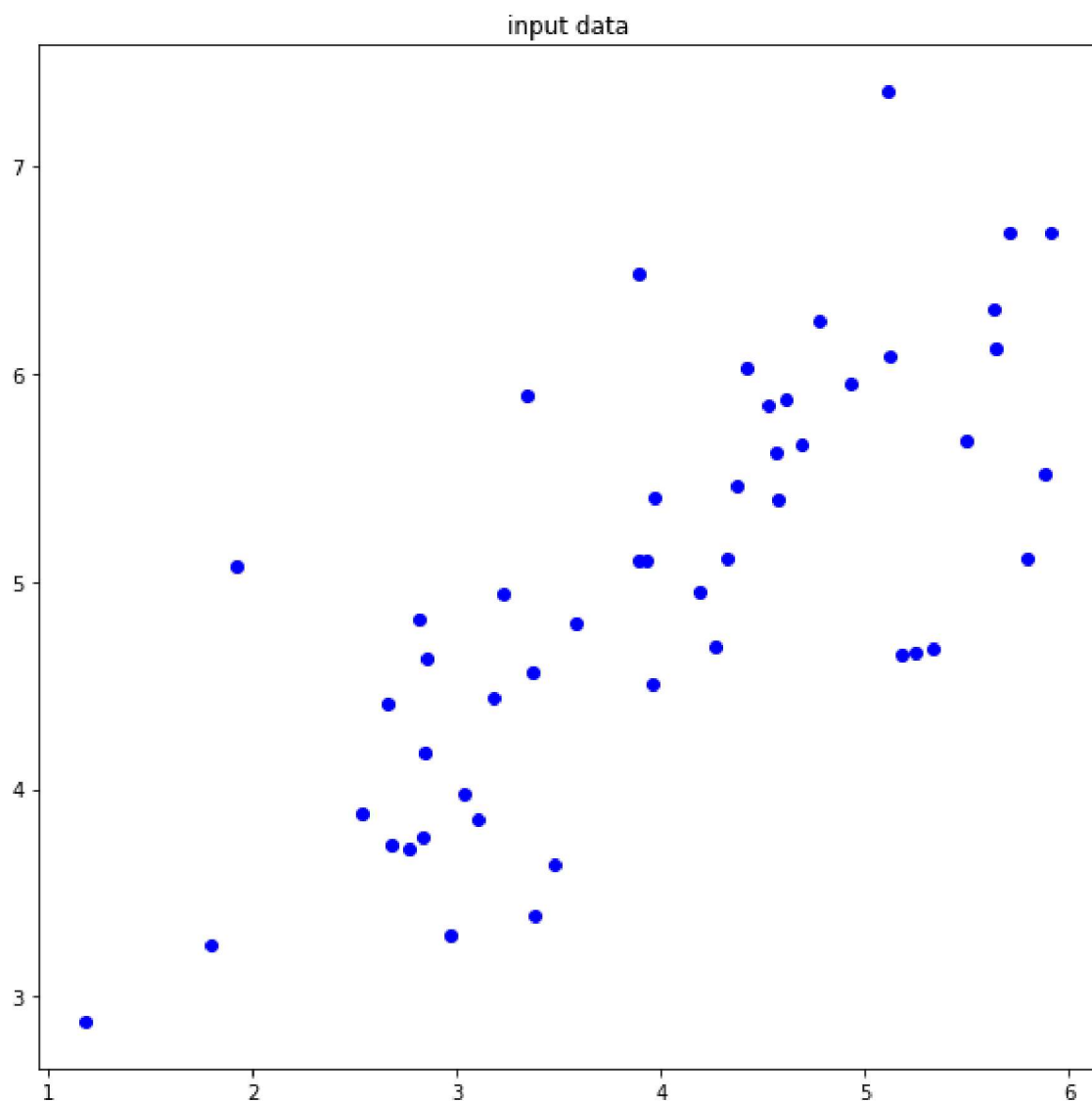
x = feature[:,0]
y = feature[:,1]

number_data = np.size(feature, 0)
number_feature = np.size(feature, 1)
# print(number_data)
# print(number_feature)
```

plot the input data

In [3]:

```
def plot_data(feature):  
    plt.figure(figsize=(8,8))  
    plt.title('input data')  
  
    x = feature[:,0]  
    y = feature[:,1]  
  
    plt.scatter(x,y,color='blue')  
  
    plt.tight_layout()  
    plt.show()  
  
plot_data(feature)
```



Normalization (Z-scoring)

In [4]:



```
def normalization(feature) :  
    x = feature[:,0]  
    y = feature[:,1]  
    mean_x = np.mean(x)  
    mean_y = np.mean(y)  
    std_x = np.std(x)  
    std_y = np.std(y)  
    #     print("mean of x =",mean_x)  
    #     print("mean of y =",mean_y)  
    #     print("std of x =",std_x)  
    #     print("std of y =",std_y)  
    x = (x-mean_x)/std_x  
    y = (y-mean_y)/std_y  
    return x, y  
  
def plot_normalization(feature) :  
    plt.figure(figsize=(8,8))  
    plt.title('data normalized by z-scoring')  
    x, y = normalization(feature)  
    plt.xlim(-2.5, 2.5)  
    plt.ylim(-2.5, 2.5)  
    plt.scatter(x,y,color='blue')  
  
    plt.tight_layout()  
    plt.show()  
# plot_normalization(feature)
```

compute covariance matrix

In [5]:



```
def compute_covariance(feature):
    x, y = normalization(feature)
    Sigma = np.cov(np.transpose(x) , np.transpose(y))
    return Sigma
compute_covariance(feature)
```

Out[5]:

```
array([[1.02040816, 0.75054082],
       [0.75054082, 1.02040816]])
```

compute principal directions

In [6]:



```
def compute_eigen(X):
    S, U = np.linalg.eig(X)
    return S, U
X = compute_covariance(feature)
compute_eigen(X)
```

Out[6]:

```
(array([1.77094898, 0.26986734]),
 array([[ 0.70710678, -0.70710678],
        [ 0.70710678,  0.70710678]]))
```

In [7]:



```
def plot_quiver(X):

    eigen_vec_1_S, eigen_vec_1_U = compute_eigen(X)[1][:,0]
    eigen_vec_2_S, eigen_vec_2_U = compute_eigen(X)[1][:,1]
    x, y = normalization(feature)

    plt.figure(figsize=(8,8))
    plt.title('principal directions')
    plt.xlim(-2.5, 2.5)
    plt.ylim(-2.5, 2.5)
    plt.scatter(x,y,color='blue')
    plt.quiver(0,0,eigen_vec_1_S,eigen_vec_1_U,angles='xy',color='red', scale=5)
    plt.quiver(0,0,eigen_vec_2_S,eigen_vec_2_U,angles='xy',color='green', scale=5)

    plt.tight_layout()
    plt.show()

# plot_quiver(X)
```

first principal axis

In [8]:



```
def plot_first_axis(X) :  
  
    eigen_vec_1_S, eigen_vec_1_U = compute_eigen(X)[1][:,0]  
    x, y = normalization(feature)  
  
    plt.figure(figsize=(8,8))  
    plt.title('first principle axis')  
    plt.scatter(x,y,color='blue')  
    plt.xlim(-2.5, 2.5)  
    plt.ylim(-2.5, 2.5)  
    inc_1 = eigen_vec_1_U / eigen_vec_1_S  
    yy = inc_1*x  
    plt.plot(x,yy, '-', color = 'red')  
  
    plt.tight_layout()  
    plt.show()  
  
# plot_first_axis(X)
```

second principal axis

In [9]:



```
def plot_second_axis(X) :  
  
    eigen_vec_2_S, eigen_vec_2_U = compute_eigen(X)[1][:,1]  
    x, y = normalization(feature)  
  
    plt.figure(figsize=(8,8))  
    plt.title('second principle axis')  
    plt.scatter(x,y,color='blue')  
    plt.xlim(-2.5, 2.5)  
    plt.ylim(-2.5, 2.5)  
    inc_2 = eigen_vec_2_U / eigen_vec_2_S  
    yy = inc_2*x  
    plt.plot(x,yy, '-', color = 'red')  
  
    plt.tight_layout()  
    plt.show()  
  
# plot_second_axis(X)
```

compute the projection of point onto the line

In [10]:



```
def compute_projection_onto_line(vector0, vector1):

    vector = np.array([vector0, vector1])

    square = np.square(np.linalg.norm(vector))
    projection = []

    x, y = normalization(feature)
    feature[:,0]=x
    feature[:,1]=y

    for i in feature:
        vd = i.dot(vector)
        projection.append([ (vector0 * vd)/square , (vector1 * vd)/square ])

    projection = np.array(projection)

    return projection
```

projcet to the first principal axis

In [11]:



```
def plot_first_projection(X) :
    eigen_vec_1_S, eigen_vec_1_U = compute_eigen(X)[1][:,0]
    x, y = normalization(feature)

    plt.figure(figsize=(8,8))
    plt.title('projection to the first principle axis')
    plt.xlim(-2.5, 2.5)
    plt.ylim(-2.5, 2.5)
    inc_1 = eigen_vec_1_U / eigen_vec_1_S
    plt.scatter(x,y,color='blue')
    yy = inc_1 * x
    plt.plot(x,yy, '-', color = 'red')

    res = compute_projection_onto_line(eigen_vec_1_S, eigen_vec_1_U)
    plt.scatter(res[:,0],res[:,1] , color='green', zorder=10)

    plt.tight_layout()
    plt.show()

# plot_first_projection(X)
```

projcet to the second principal axis

In [12]:



```
def plot_second_projection(X) :
    eigen_vec_2_S, eigen_vec_2_U = compute_eigen(X)[1][:,1]
    x, y = normalization(feature)

    plt.figure(figsize=(8,8))
    plt.title('projection to the second principle axis')
    plt.xlim(-2.5, 2.5)
    plt.ylim(-2.5, 2.5)
    inc_2 = eigen_vec_2_U / eigen_vec_2_S
    plt.scatter(x,y,color='blue')
    yy = inc_2 * x
    plt.plot(x,yy, '-', color = 'red')

    res = compute_projection_onto_line(eigen_vec_2_S, eigen_vec_2_U)
    plt.scatter(res[:,0],res[:,1] , color='green', zorder=10)

    plt.tight_layout()
    plt.show()

# plot_second_projection(X)
```

connect original data to the projection onto the first principal axis

In [13]:



```
def plot_first_projection_axis(X) :
    eigen_vec_1_S, eigen_vec_1_U = compute_eigen(X)[1][:,0]
    x, y = normalization(feature)
    feature[:,0]=x
    feature[:,1]=y

    plt.figure(figsize=(8,8))
    plt.title('projection to the first principle axis')
    plt.xlim(-2.5, 2.5)
    plt.ylim(-2.5, 2.5)
    inc_1 = eigen_vec_1_U / eigen_vec_1_S
    plt.scatter(x,y,color='blue',zorder=8)
    yy = inc_1 * x
    plt.plot(x,yy, '-', color = 'red')

    res = compute_projection_onto_line(eigen_vec_1_S, eigen_vec_1_U)
    plt.scatter(res[:,0],res[:,1] , color='green', zorder=10)

    for i in range(0, number_data):
        plt.plot([feature[i][0],res[i][0]],[feature[i][1],res[i][1]], '-',color='black')

    plt.tight_layout()
    plt.show()

# plot_first_projection_axis(X)
```

connect original data to the projection onto the second principal axis

In [14]:



```
def plot_second_projection_axis(X) :
    eigen_vec_2_S, eigen_vec_2_U = compute_eigen(X)[1][:,1]
    x, y = normalization(feature)
    feature[:,0]=x
    feature[:,1]=y

    plt.figure(figsize=(8,8))
    plt.title('projection to the second principle axis')
    plt.xlim(-2.5, 2.5)
    plt.ylim(-2.5, 2.5)
    inc_2 = eigen_vec_2_U / eigen_vec_2_S
    plt.scatter(x,y,color='blue',zorder=8)
    yy = inc_2 * x
    plt.plot(x,yy, '-', color = 'red')

    res = compute_projection_onto_line(eigen_vec_2_S, eigen_vec_2_U)
    plt.scatter(res[:,0],res[:,1] , color='green', zorder=10)

    for i in range(0, number_data):
        plt.plot([feature[i][0],res[i][0]],[feature[i][1],res[i][1]], '-',color='black')

    plt.tight_layout()
    plt.show()

# plot_second_projection_axis(X)
```

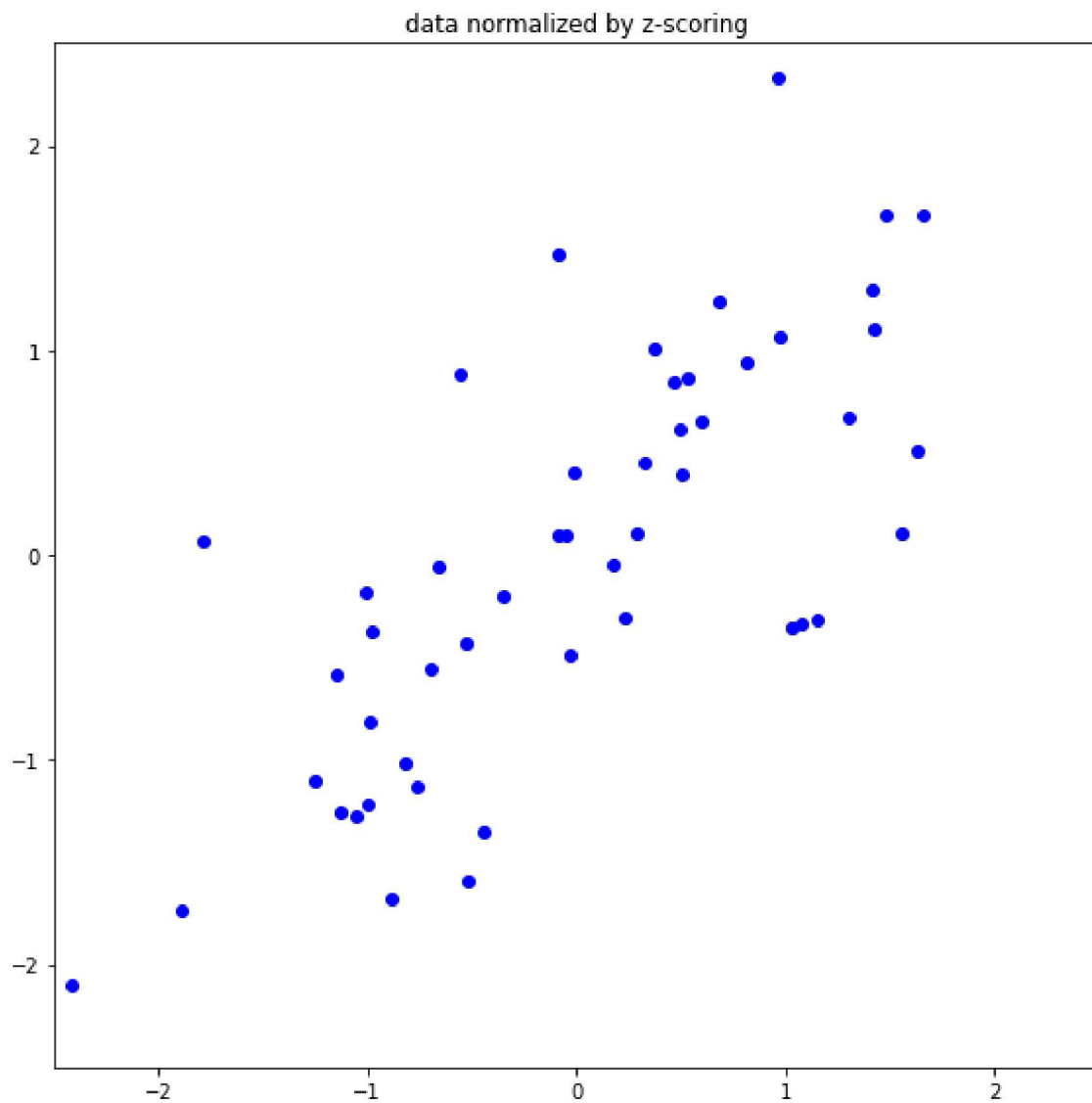
Note that the title in the above figure is wrong. It should be "second principle axis" instead of "first principle axis"

results

1. plot the input data after the normalization using Z-scoring in blue

In [15]:

```
plot_normalization(feature)
```

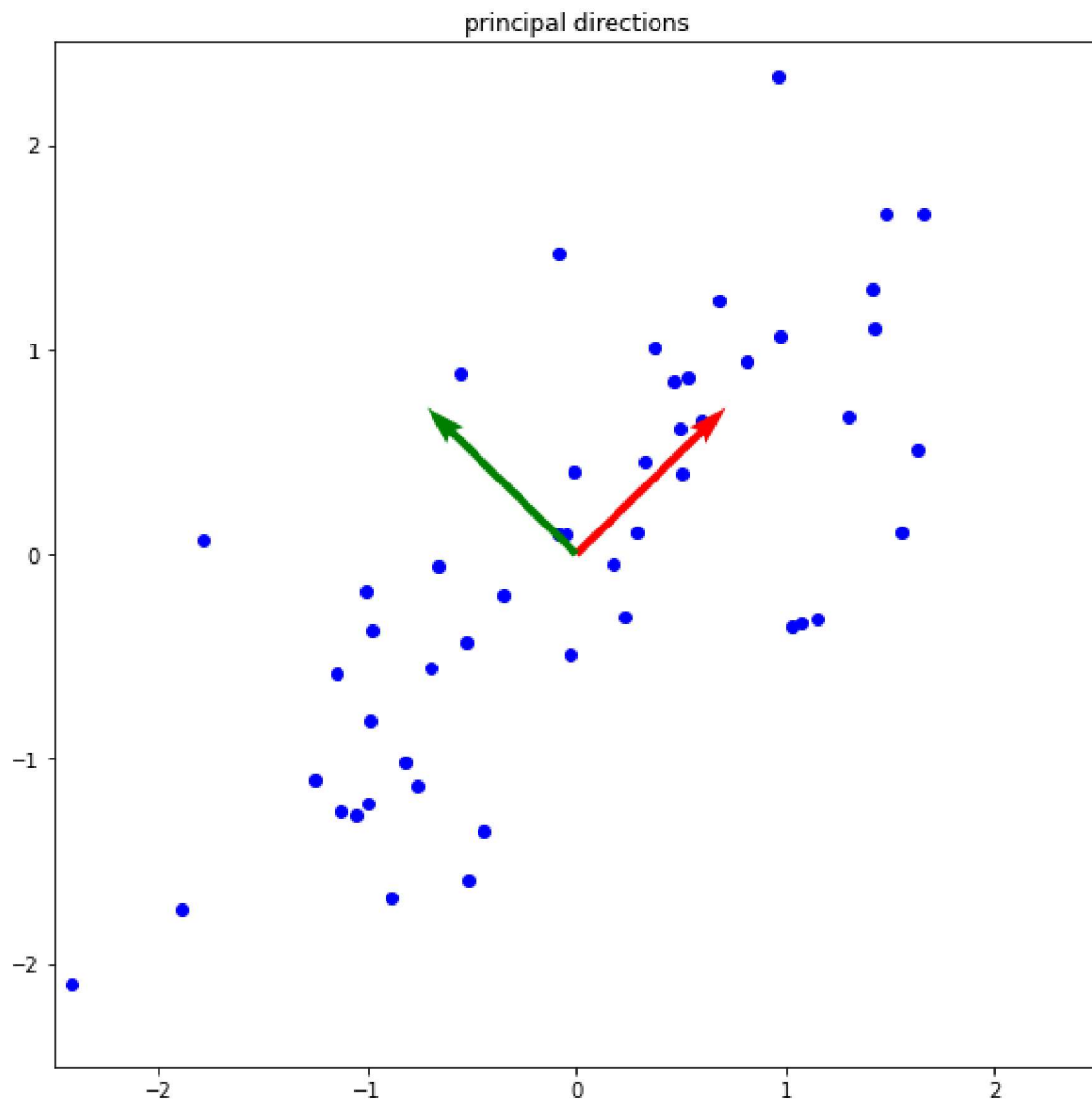


2. plot the first principal component in red and the second principal components in green on the normalized data in blue

In [16]:



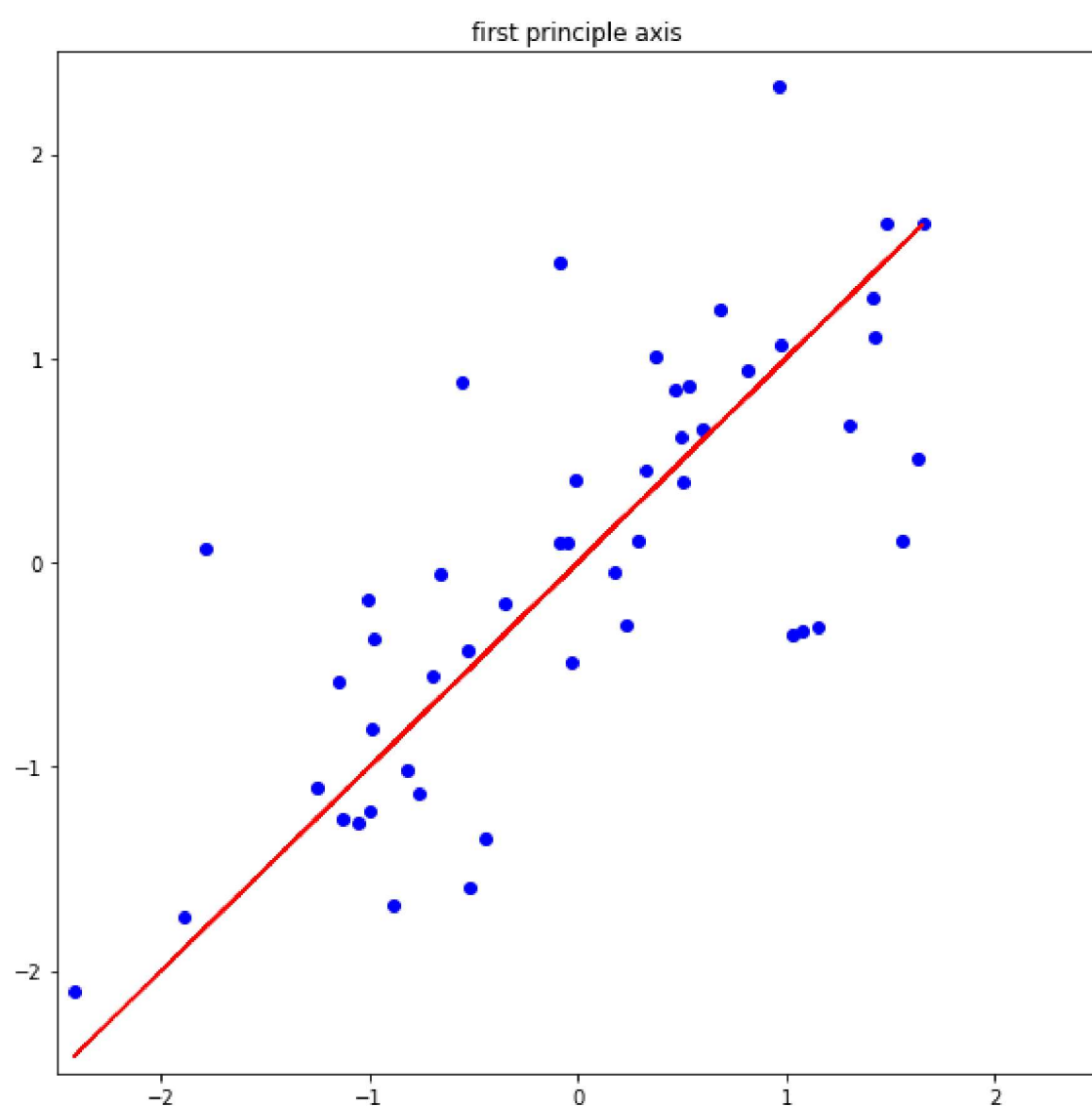
```
plot_quiver(X)
```



3. plot the first principal axis in red on the normalized data in blue

In [17]:

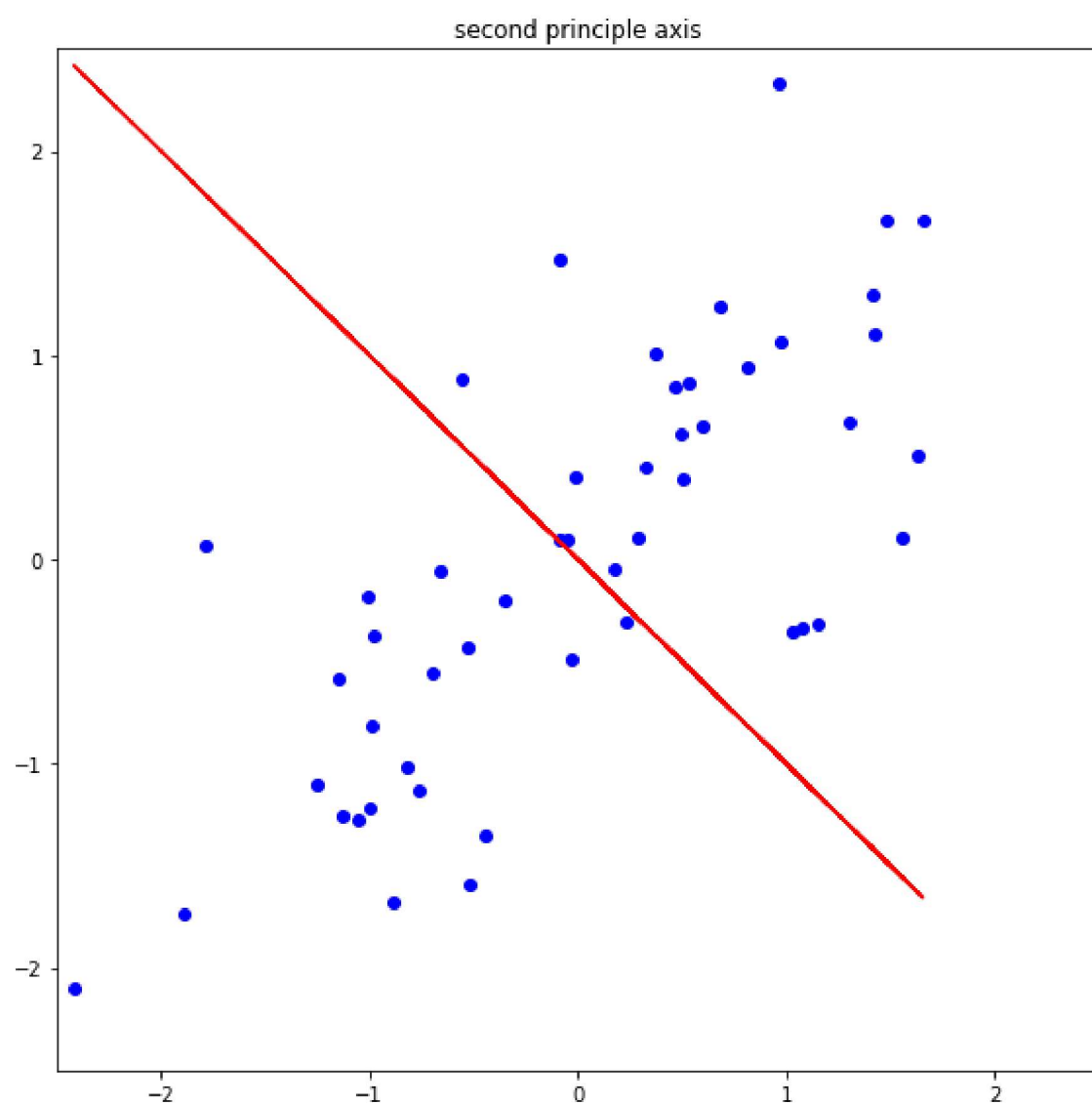
```
plot_first_axis(X)
```



4. plot the second principal axis in red on the normalized data in blue

In [18]:

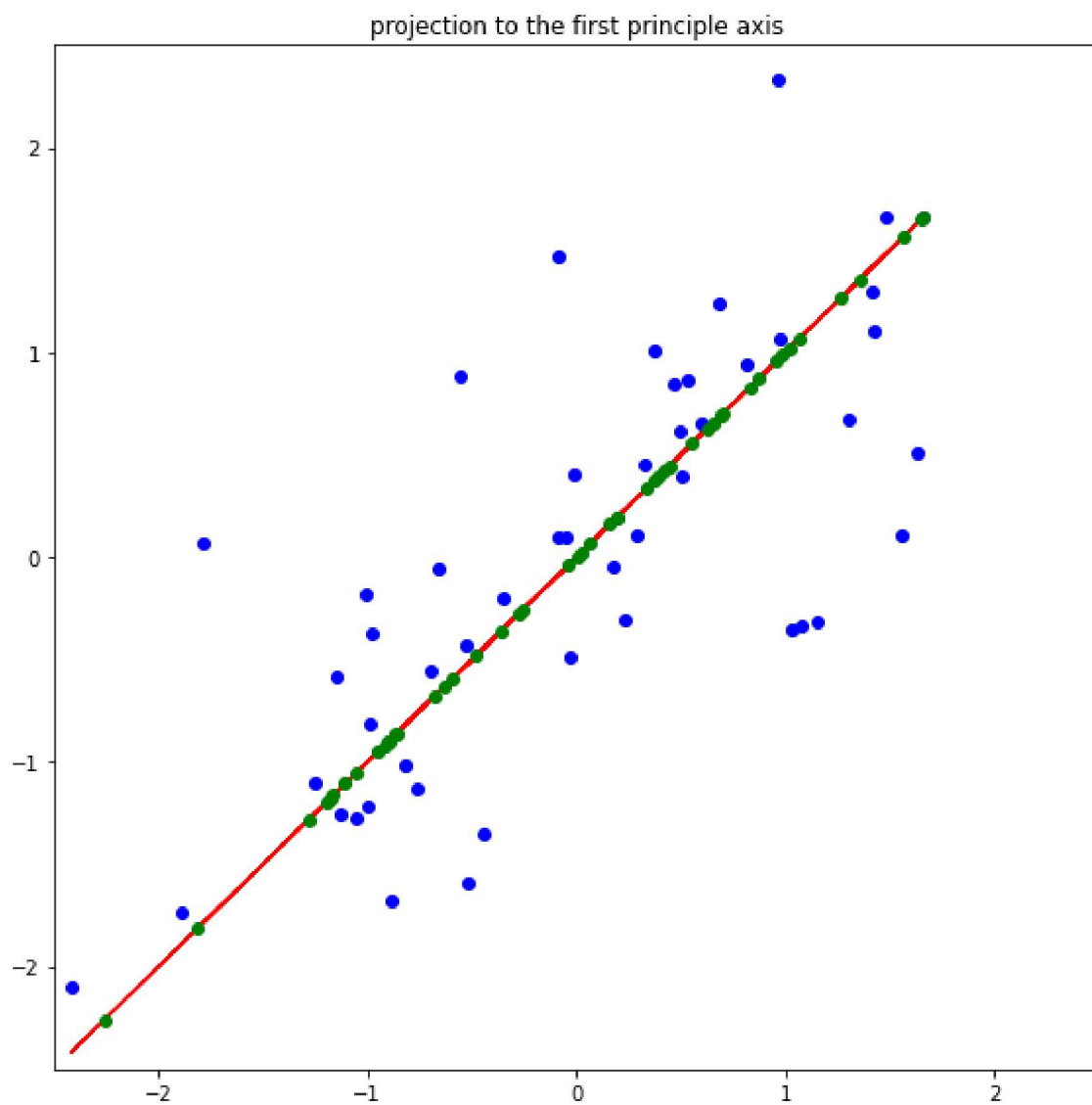
```
plot_second_axis(X)
```



5. plot the projection of data in green onto the first principal axis in red

In [19]:

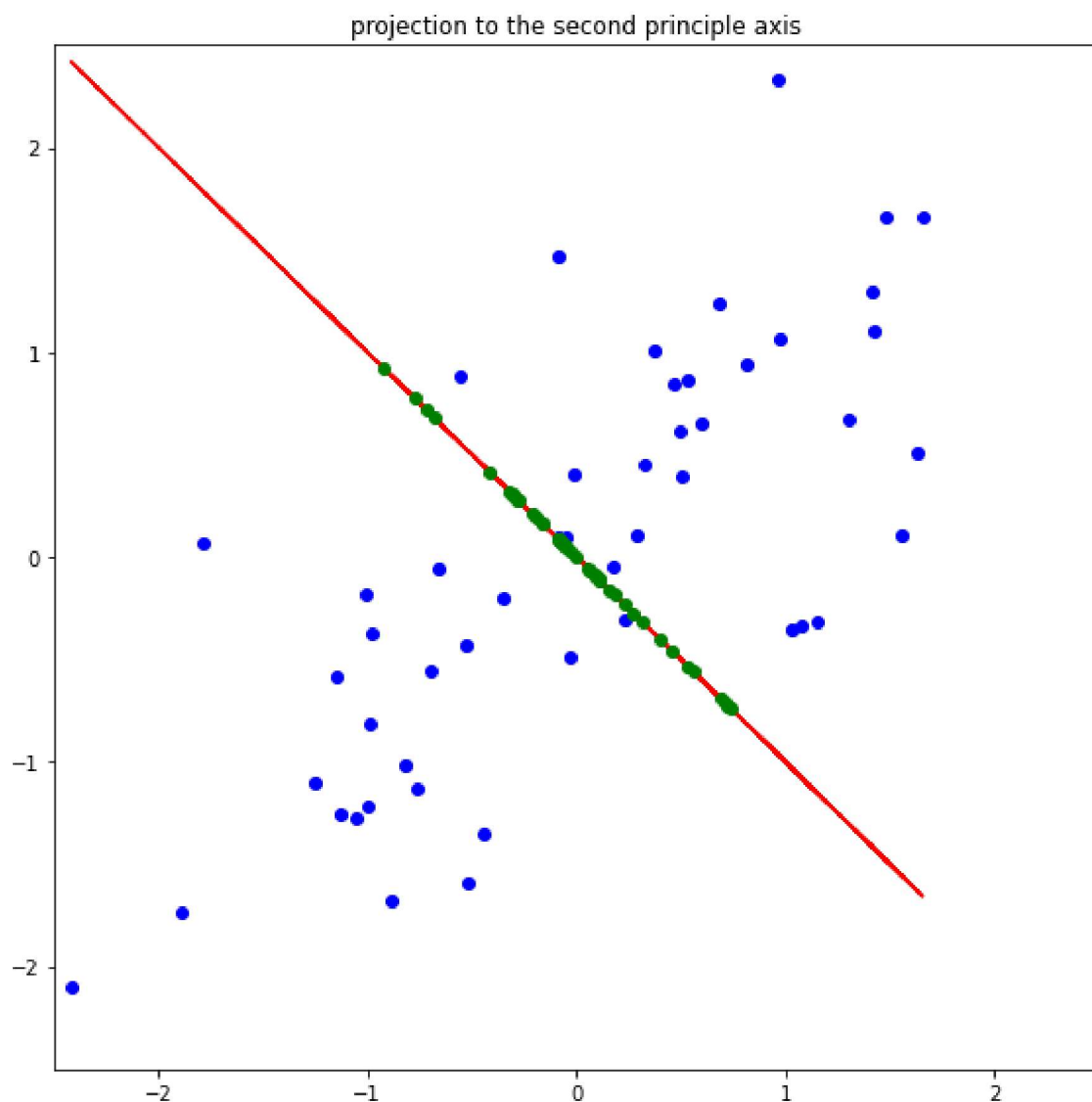
```
plot_first_projection(X)
```



6. plot the projection of data in green onto the second principal axis in red

In [20]:

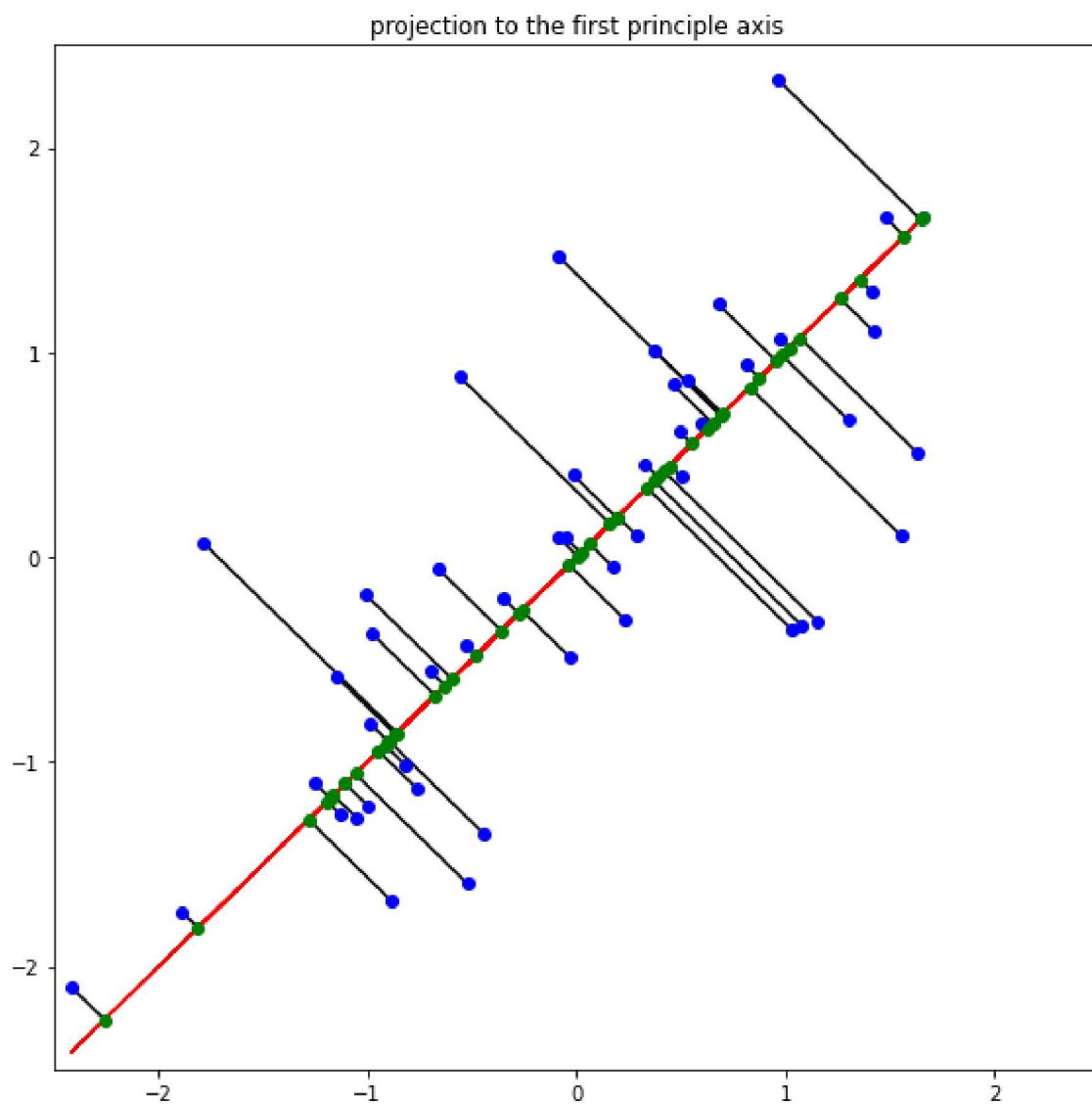
```
plot_second_projection(X)
```



7. plot the projection line in grey onto the first principal axis

In [21]:

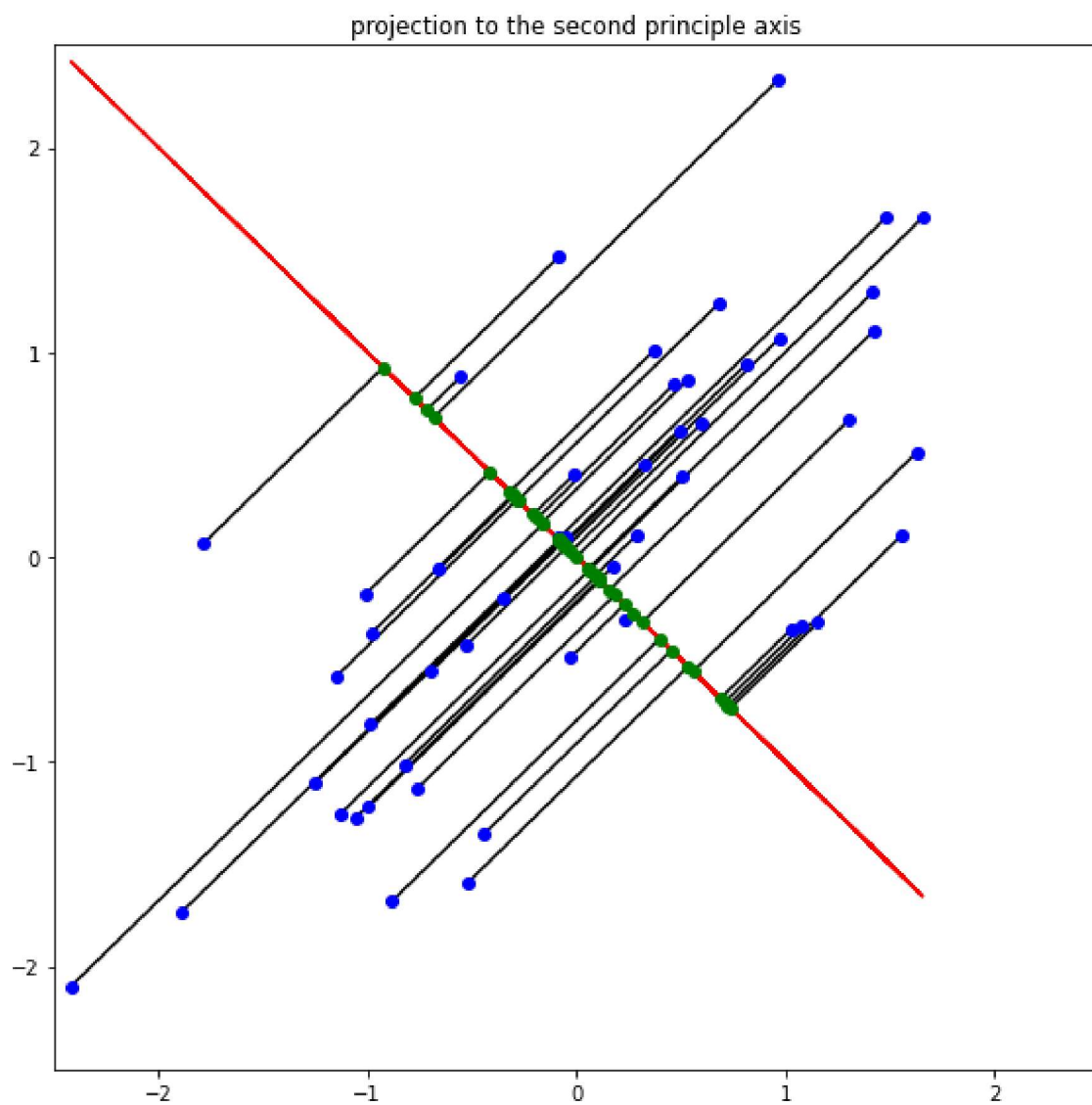
```
plot_first_projection_axis(X)
```



8. plot the projection line in grey onto the second principal axis

In [22]:

```
plot_second_projection_axis(X)
```



Note that the title in the above figure is wrong. It should be "second principle axis" instead of "first principle axis"