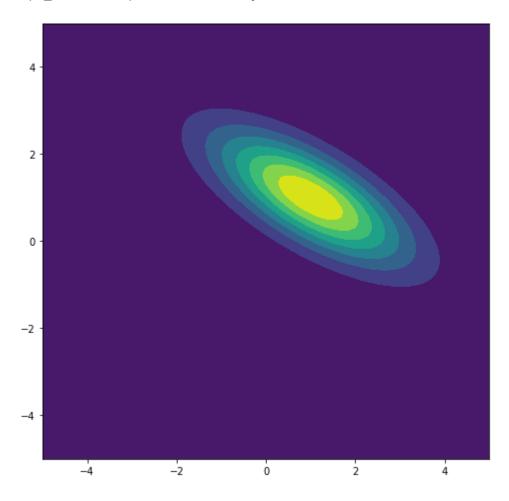
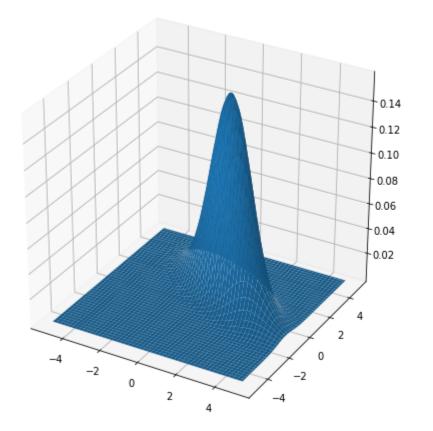
1 - Visualizing the Multivariate Gaussian (Normal) Probability Density Function

```
In [172]: import numpy as np
            import matplotlib.pyplot as plt
            from scipy.stats import multivariate normal
            from mpl toolkits.mplot3d import Axes3D
            Х
                         = np.linspace(-5,5,100)
                      = np.linspace(-5,5,100)
= np.meshgrid(x,y)
            X,Y
                        = np.meshgrid(x,y)
                         = np.zeros(X.shape + (2,))
            pos
            pos[:,:,0] = X
            pos[:,:,1] = Y
                       = [1,1]
            mean
            COV
                         = [[2,1],[-1,1]]
            var = [[2,1],[-1,1]]
var = multivariate_normal(mean, cov)
fig0 = plt.figure(figsize=(8,8))
c_fig = fig0.add_subplot(111)
            c fig.contourf(X,Y,var.pdf(pos))
            fig1
                         = plt.figure(figsize=(8,8))
            surf fig = fig1.add subplot(111,projection='3d')
            surf_fig.plot_surface(X,Y,var.pdf(pos))
```

Out[172]: <mpl_toolkits.mplot3d.art3d.Poly3DCollection at 0x269a86db6a0>

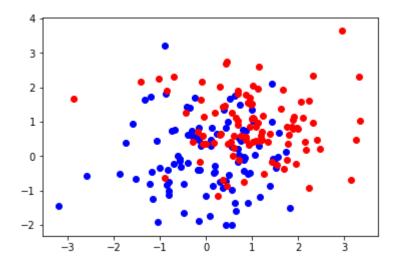




2 - Generate Synthetic Data for Classification

```
In [171]:
          import numpy as np
          import matplotlib.pyplot as plt
          from scipy.stats import multivariate normal
          # mean and cov for the first dataset
          mean1 = [0,0]
          cov1 = [[1,0.1],[0.1,1]]
          # mean and cov for the second dataset
          mean2 = [1,1]
          cov2 = [[1,-0.1],[-0.1,1]]
          # initializing sample points
          data1
                           = np.random.multivariate normal(mean1,cov1,100)
          data2
                           = np.random.multivariate normal(mean2,cov2,100)
          # Plotting
          fig_gen_syn_data = plt.figure()
          plot gsd
                           = fig gen syn data.add subplot(111)
          plot gsd.scatter(data1[:,0],data1[:,1],c='blue')
          plot_gsd.scatter(data2[:,0],data2[:,1],c='red')
```

Out[171]: <matplotlib.collections.PathCollection at 0x269a8766898>

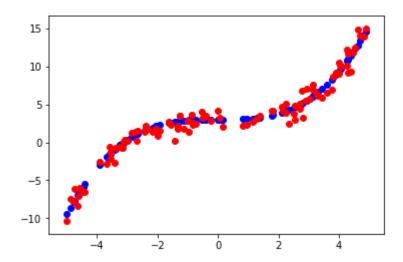


3 - Generate Synthetic Data for Regression

```
In [170]: import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import multivariate_normal

uniform_var = np.random.uniform(-5,5,100)
target_val = 0.1 * (uniform_var**3) + 3
noise = np.random.normal(size=100)
noisy_obs = target_val + noise
# plotting
fig_noisy_data = plt.figure()
plot = fig_noisy_data.add_subplot(111)
plot.scatter(uniform_var,target_val,c='blue')
plot.scatter(uniform_var,noisy_obs,c='red')
```

Out[170]: <matplotlib.collections.PathCollection at 0x269a9ff3780>



4 - Standardizing Data

In []:

```
In [169]:
           import numpy as np
           import matplotlib.pyplot as plt
           from scipy.stats import multivariate_normal
           mean
                      = [2,5]
           cov
                      = [[1,-0.1],[-0.1,1]]
                      = np.random.multivariate_normal(mean,cov,size=10)
           Χ
           X hat
                      = np.empty(X.shape)
           X_{\text{hat}}[:,0] = (X[:,0]-np.mean(X[:,0]))/np.std(X[:,0])
           X_{\text{hat}}[:,1] = (X[:,1]-np.mean(X[:,1]))/np.std(X[:,1])
                      = plt.figure()
           fig1
                      = fig1.add_subplot(111)
           plot1
           plot1.scatter(X[:,0],X[:,1], c='blue')
           fig2
                      = plt.figure()
                      = fig2.add_subplot(111)
           plot2
           plot2.scatter(X_hat[:,0],X_hat[:,1], c='red')
```

Out[169]: <matplotlib.collections.PathCollection at 0x269aa084668>

