Question 3 - Part B

Probabilistic Supervised Learning - Naive Bayes(Binomial): Create a dataset from the sample given to you(e.g. "Titanic, Play Tennis Probability", "Shopper Buying Probability" etc.). Perform the necessary pre-processing steps such as encoding. Train the model using Naive Bayes Classifier for Binomial predictions. Give new test data and predict the classification output. Handcode the classification probability and compare with the model output. Analyze and write the inference.

Q3

Kagle Titanic dataset

```
In [61]: import pandas as pd
#Import dataset
df = pd.read_csv('titanic.csv')
df.head()
```

Out[61]:

```
Siblings/Spouses
                                                                     Parents/Children
   Survived Pclass
                           Name
                                     Sex Age
                                                                                          Fare
                                                           Aboard
                                                                              Aboard
                       Mr. Owen
0
          0
                                                                 1
                  3
                           Harris
                                    male 22.0
                                                                                    0
                                                                                        7.2500
                          Braund
                        Mrs. John
                          Bradley
                        (Florence
                                                                 1
                                                                                    0 71.2833
           1
                                   female 38.0
                           Briggs
                          Thayer)
                           Cum...
                       Miss. Laina
2
          1
                                   female 26.0
                                                                 0
                                                                                        7.9250
                       Heikkinen
                            Mrs.
                         Jacques
3
                                                                 1
          1
                                                                                    0 53.1000
                       Heath (Lily
                                  female 35.0
                       May Peel)
                          Futrelle
                      Mr. William
          0
                                                                 0
                                                                                        8.0500
4
                                    male 35.0
                      Henry Allen
```

```
In [62]: #Drop extra things
    df.dropna()
    df.drop(['Name',],inplace=True,axis=1)
    df.head()
```

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```
Out[62]:
                                               Siblings/Spouses
                                                                     Parents/Children
             Survived Pclass
                               Sex Age
                                                                                       Fare
                                                       Aboard
                                                                             Aboard
          0
                   0
                          3
                              male 22.0
                                                            1
                                                                                  0
                                                                                     7.2500
          1
                   1
                          1 female
                                    38.0
                                                            1
                                                                                  0 71.2833
          2
                   1
                          3 female
                                   26.0
                                                            0
                                                                                  0
                                                                                     7.9250
                                   35.0
                                                                                    53.1000
          3
                   1
                          1
                             female
                                                            1
          4
                   0
                          3
                              male 35.0
                                                            0
                                                                                  \cap
                                                                                     8.0500
In [63]:
          #Encoding text to numbers
          from sklearn.preprocessing import LabelEncoder,StandardScaler
          encoder = LabelEncoder()
          df['Sex']=encoder.fit transform(df['Sex'])
In [64]:
         #Using scaler to standardize the mean and variance
          scaler = StandardScaler()
          df[['Fare','Age']] = scaler.fit_transform(df[['Fare','Age']])
          df.head()
Out[64]:
                                               Siblings/Spouses
                                                                   Parents/Children
             Survived Pclass Sex
                                      Age
                                                                                       Fare
                                                                           Aboard
                                                       Aboard
          0
                   0
                          3
                               1 -0.529366
                                                            1
                                                                                0 -0.503586
          1
                   1
                          1
                                  0.604265
                                                                                    0.783412
                                                            1
          2
                   1
                          3
                               0 -0.245958
                                                            0
                                                                                0 -0.490020
          3
                                  0.391709
                                                                                    0.417948
          4
                   0
                          3
                                  0.391709
                                                            0
                                                                                  -0.487507
In [65]: #Training model
          from sklearn.naive bayes import GaussianNB
          from sklearn.model_selection import train_test_split
          #Train test split
          x_train,x_test,y_train,y_test=train_test_split(df.drop(['Survived'],axis=1),df['
          model =GaussianNB()
          model.fit(x_train,y_train)
          #Listing classes
          model.classes
Out[65]: array([0, 1], dtype=int64)
In [66]:
         #Prediction and calculate accuracy
          y_pred=model.predict(x_test)
          from sklearn.metrics import confusion_matrix,accuracy_score,f1_score
          acc = accuracy_score(y_pred,y_test)
          f1 = f1_score(y_pred,y_test)
          print("Accuracy : ",acc,"\nF1 Score : ",f1)
          Accuracy: 0.7865168539325843
          F1 Score: 0.7205882352941175
```

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```
In [67]: #Draw confusion matrix
    print(y_test.shape)
    labels = [0,1]
    cm=confusion_matrix(y_pred,y_test,labels=labels)

import seaborn as sns
    sns.heatmap(cm,annot=True,cmap='Blues')

(178,)
```

Out[67]: <AxesSubplot: >

