# KNN implementation on Gene Expression Data Set (Generalized)

## 60% train data, 40% test data

# Library

```
In [2]:
```

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import math
```

### **Data Set**

```
In [3]:
```

```
df = pd.read_csv('gene_expression.csv')
df
```

### Out[3]:

	Gene One	Gene Two	<b>Cancer Present</b>
0	4.3	3.9	1
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
2995	5.0	6.5	1
2996	3.4	6.6	0
2997	2.7	6.5	0
2998	3.3	5.6	0
2999	4.6	8.2	0

### 3000 rows × 3 columns

3

```
In [4]:
df.columns
Out[4]:
Index(['Gene One', 'Gene Two', 'Cancer Present'], dtype='object')
In [5]:
n = len(df.columns)
n
Out[5]:
```

```
ո [6]:
df[df.columns[n-1]].value counts()
Out[6]:
     1500
0
     1500
Name: Cancer Present, dtype: int64
In [7]:
df[df.columns[n-1]].value counts().index
Out[7]:
Int64Index([1, 0], dtype='int64')
In [8]:
df[df.columns[n-1]].nunique()
Out[8]:
2
In [9]:
custom palette = sns.color palette("Set1",10)
sns.palplot(custom palette)
```

# https://www.codecademy.com/article/seaborn-design-ii

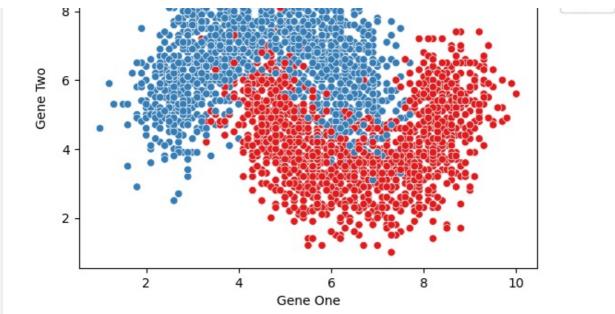
### In [10]:

```
custom_palette = sns.color_palette("Set1", df[df.columns[2]].nunique()+1)
color dict = dict()
markers dict = dict()
j = 0
for i in df[df.columns[2]].value counts().index:
   color_dict[i] = custom_palette[j]
   markers_dict[i] = 'o'
    j = j + 1
color dict['Test Point'] = custom palette[2]
markers dict['Test Point'] = 'X'
# print(color dict)
# print(markers dict)
sns.scatterplot(x = df[df.columns[0]], y = df[df.columns[1]], hue=df[df.columns[2]], palett
e=color dict,style=df[df.columns[2]],markers=markers dict)
plt.title('Full Data Points')
plt.legend(loc=(1.05, 0.75))
```

### Out[10]:

<matplotlib.legend.Legend at 0x1622da71660>

# Full Data Points



# **Determing the value of K**

```
In [11]:
len(df)
Out[11]:
3000
In [12]:
k = math.floor(math.sqrt(len(df)))
if k\%2==0:
    k = k + 1
print(k)
55
```

# **Train Test split**

df.head(no of train data)

```
In [13]:
train percentage = 60
test_percentage = 100 - train_percentage
print('Train Percentage :', train percentage)
print('Test Percentage :',test_percentage)
Train Percentage: 60
Test Percentage: 40
In [14]:
no of train data = math.ceil((train percentage * len(df)) / 100)
print('No of train data :',no_of_train_data)
no of test data = len(df) - no of train data
print('No of test data', no_of_test_data)
No of train data: 1800
No of test data 1200
In [15]:
```

Out[15]:

	Gene One	Gene Two	<b>Cancer Present</b>
0	4.3	3.9	1
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
1795	2.9	7.3	0
1796	6.1	3.3	1
1797	6.5	8.0	0
1798	2.5	5.5	0
1799	2.6	6.0	0

1800 rows × 3 columns

```
In [16]:
```

```
df_train = df.head(no_of_train_data)
df_train
```

Out[16]:

	Gene One	Gene Two	<b>Cancer Present</b>
0	4.3	3.9	1
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
1795	2.9	7.3	0
1796	6.1	3.3	1
1797	6.5	8.0	0
1798	2.5	5.5	0
1799	2.6	6.0	0

1800 rows × 3 columns

```
In [17]:
```

```
df.tail(no_of_test_data)
```

Out[17]:

	Gene One	Gene Two	<b>Cancer Present</b>
1800	8.1	3.4	1
1801	7.7	4.1	1
1802	5.1	4.5	1
1803	4.3	9.6	0
1804	7.7	5.4	1

2995	Gene One	Gene Two	Cancer Present
2996	3.4	6.6	0
2997	2.7	6.5	0
2998	3.3	5.6	0
2999	4.6	8.2	0

### 1200 rows × 3 columns

```
In [18]:
```

```
df_test = df.tail(no_of_test_data)
df_test
```

### Out[18]:

	Gene One	Gene Two	<b>Cancer Present</b>
1800	8.1	3.4	1
1801	7.7	4.1	1
1802	5.1	4.5	1
1803	4.3	9.6	0
1804	7.7	5.4	1
2995	5.0	6.5	1
2996	3.4	6.6	0
2997	2.7	6.5	0
2998	3.3	5.6	0
2999	4.6	8.2	0

### 1200 rows × 3 columns

### In [19]:

```
df_test = df_test.reset_index()
df_test
```

### Out[19]:

	index	Gene One	Gene Two	Cancer Present
0	1800	8.1	3.4	1
1	1801	7.7	4.1	1
2	1802	5.1	4.5	1
3	1803	4.3	9.6	0
4	1804	7.7	5.4	1
1195	2995	5.0	6.5	1
1196	2996	3.4	6.6	0
1197	2997	2.7	6.5	0
1198	2998	3.3	5.6	0
1199	2999	4.6	8.2	0

### 1200 rows × 4 columns

```
df_test = df_test.drop('index',axis=1)
df_test
```

Out[20]:

	Gene One	Gene Two	Cancer Present
0	8.1	3.4	1
1	7.7	4.1	1
2	5.1	4.5	1
3	4.3	9.6	0
4	7.7	5.4	1
1195	5.0	6.5	1
1196	3.4	6.6	0
1197	2.7	6.5	0
1198	3.3	5.6	0
1199	4.6	8.2	0

### 1200 rows × 3 columns

```
In [21]:
```

```
df_temp = df.copy()
df_temp
```

Out[21]:

	Gene One	Gene Two	<b>Cancer Present</b>
0	4.3	3.9	1
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
2995	5.0	6.5	1
2996	3.4	6.6	0
2997	2.7	6.5	0
2998	3.3	5.6	0
2999	4.6	8.2	0

### 3000 rows × 3 columns

```
In [22]:
```

```
df_temp['Cancer Present'][no_of_train_data:] = ['Test Point'] * no_of_test_data
df_temp

C:\Users\HP\AppData\Local\Temp\ipykernel_4124\2074753982.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_g
uide/indexing.html#returning-a-view-versus-a-copy
    df_temp['Cancer Present'][no_of_train_data:] = ['Test Point'] * no_of_test_data
```

Out[22]:

	Gene One	Gene Two	Cancer Present
-0	4.3	3.9	
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
2995	5.0	6.5	Test Point
2996	3.4	6.6	Test Point
2997	2.7	6.5	Test Point
2998	3.3	5.6	Test Point
2999	4.6	8.2	Test Point

### 3000 rows × 3 columns

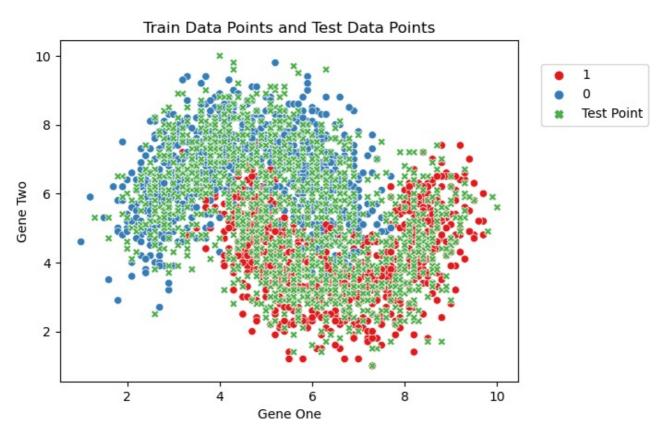
### In [23]:

```
sns.scatterplot(x = df_temp[df_temp.columns[0]],y = df_temp[df_temp.columns[1]],hue=df_t
emp[df_temp.columns[2]],palette=color_dict,style=df_temp[df_temp.columns[2]],markers=mar
kers_dict)
plt.title('Train Data Points and Test Data Points')
plt.legend(loc=(1.05,0.75))

# hue without palette : sns will provide default color for each group or class in df_temp
['Cancer Present]
# huw with palette : sns will provide color we want for each group or class in df_temp['C
ancer Present]
# style without markers : sns will provide default shape for each group or class in df_te
mp['Cancer Present]
# style with markers : sns will provide shape we want for each group or class in df_temp['Cancer Present]
```

### Out[23]:

<matplotlib.legend.Legend at 0x1623352bfd0>



### Calculating Fuclean Distance from Test point to Train point - sorting it

# ascending order and then finding the nearest neighbor

```
In [24]:
```

df\_train

Out[24]:

	Gene One	Gene Two	<b>Cancer Present</b>
0	4.3	3.9	1
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
1795	2.9	7.3	0
1796	6.1	3.3	1
1797	6.5	8.0	0
1798	2.5	5.5	0
1799	2.6	6.0	0

### 1800 rows × 3 columns

### In [25]:

df\_test

Out[25]:

	Gene One	Gene Two	<b>Cancer Present</b>
0	8.1	3.4	1
1	7.7	4.1	1
2	5.1	4.5	1
3	4.3	9.6	0
4	7.7	5.4	1
1195	5.0	6.5	1
1196	3.4	6.6	0
1197	2.7	6.5	0
1198	3.3	5.6	0
1199	4.6	8.2	0

### 1200 rows × 3 columns

### In [26]:

```
df_test.iloc[0]
```

### Out[26]:

Gene One 8.1
Gene Two 3.4
Cancer Present 1.0
Name: 0, dtype: float64

### In [27]:

```
df train.iloc[0][0]
Out[27]:
4.3
In [28]:
df train[df.columns[n-1]][0]
Out[28]:
1
In [29]:
df train.columns
Out[29]:
Index(['Gene One', 'Gene Two', 'Cancer Present'], dtype='object')
In [30]:
# distance list = list()
# class name list = list()
# total distance = 0
# # calculating euclidean distance from test data to train data
# for i in range(len(df test)):
#
      for j in range(len(df train)):
#
          for c in range (n-1):
#
              distance = (df_test.iloc[i][c] - df_train.iloc[j][c])**2
#
              total distance = total distance + distance
#
          total distance = math.sqrt(total distance)
#
          distance list.append((df train[df train.columns[n-1]][j],total distance))
#
          total distance = 0
#
      # sorting all those distances
#
      for ii in range(len(distance list)):
#
          for jj in range(ii+1,len(distance list)):
#
              if distance list[jj][1] < distance list[ii][1]:</pre>
#
                   temp = distance list[ii]
#
                   distance_list[ii] = distance_list[jj]
#
                   distance \ list[jj] = temp
#
      # selecting first 'k' points and then counting the number of classes
#
      count = dict()
#
      for ii in range(k):
#
          if distance list[ii][0] not in count:
#
              count[distance list[ii][0]] = 1
#
          else:
#
              count[distance list[ii][0]] = count[distance list[ii][0]] + 1
#
      # finding out the most nearest class
      min = 0
#
#
      for ii in count:
#
          if count[ii] > min:
#
              class name = ii
#
              min = count[ii]
      class name list.append(class name)
      distance list = list()
# print(class name list)
```

This code takes a lot of time for giving the output since we are using nested loop here and also used bubble sort inside the first loop which consumed too much time. and then again we had to select first k points with smallest distance using one loop and find number of classes with another loop.

so too much use of loop made this code too much slower.

min = count[ii]
class name list.append(class name)

print(class name list)

So here i have optimized this code with eliminating the bubble sort.

here , after i got the distance list , i only took the first k points with smallest distance without sorting them and everytime i took k(i) smallest point , i counted the class number, the point belongs to , too in dicitonary.after that I just found the most nearest neghbor

```
In [31]:
class name list = list()
# calculating euclidean distance from test data to train data
for i in range(len(df test)):
    distance list = list()
    for j in range(len(df_train)):
        distance = np.linalg.norm(df test.iloc[i]-df train.iloc[j])
        distance list.append((df train[df train.columns[n-1]][j], distance))
    # print(distance list)
    # selecting first 'k' points with smallest distance without sorting and then counting
the number of classes
   minimum = None
   checked = dict()
    count = dict()
   min list = list()
    for ii in range(k):
        for jj in range(len(distance list)):
            if jj not in checked :
                if minimum is None :
                    minimum = distance list[jj]
                elif distance_list[jj][1] < minimum[1]:</pre>
                    minimum = distance list[jj]
                    index = jj
        min_list.append(minimum)
        if minimum[0] not in count:
            count[minimum[0]] = 1
        else:
            count[minimum[0]] = count[minimum[0]] + 1
        checked[index] = 1
        minimum = None
    # print(min list)
    # print(count)
    # finding out the most nearest class
    min = 0
    for ii in count:
        if count[ii] > min:
            class name = ii
```

```
1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1,
1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1,
0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0,
1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1,
  1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1,
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  1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1,
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  0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1,
  0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1,
  1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1,
0, 0, 1,
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1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1,
0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1,
0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0,
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        0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
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        0, 0, 0, 1,
  0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1,
0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0]
```

### In [32]:

```
print(len(class_name_list))
```

1200

### In [33]:

```
df_temp2 = df.copy()
df_temp2
```

### Out[33]:

	Gene One	Gene Two	<b>Cancer Present</b>
0	4.3	3.9	1
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
		•••	
2995	5.0	6.5	1
2996	3.4	6.6	0
2997	2.7	6.5	0
2998	3.3	5.6	0
2999	4.6	8.2	0

### 3000 rows × 3 columns

In [34]:

```
df_temp2[df_temp2.columns[n-1]][no_of_train_data:] = class_name_list
df_temp2
C:\Users\HP\AppData\Local\Temp\ipykernel_4124\2165292902.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_g
uide/indexing.html#returning-a-view-versus-a-copy
df_temp2[df_temp2.columns[n-1]][no_of_train_data:] = class_name_list
```

### Out[34]:

	Gene One	Gene Two	Cancer Present
0	4.3	3.9	1
1	2.5	6.3	0
2	5.7	3.9	1
3	6.1	6.2	0
4	7.4	3.4	1
2995	5.0	6.5	1
2996	3.4	6.6	0
2997	2.7	6.5	0
2998	3.3	5.6	0
2999	4.6	8.2	0

### 3000 rows × 3 columns

### In [35]:

```
color_dict
```

### Out[35]:

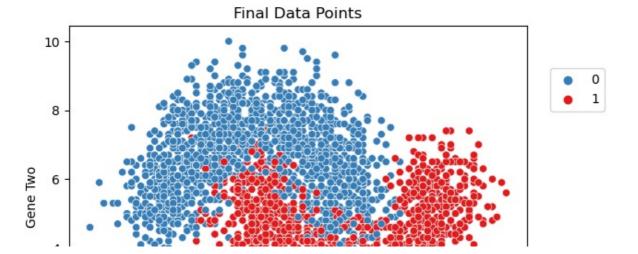
```
{1: (0.8941176470588236, 0.10196078431372549, 0.10980392156862745), 0: (0.21568627450980393, 0.49411764705882355, 0.7215686274509804), 'Test Point': (0.30196078431372547, 0.6862745098039216, 0.2901960784313726)}
```

### In [36]:

```
sns.scatterplot(x = df_temp2[df_temp2.columns[0]],y = df_temp2[df_temp2.columns[1]],hue=d
f_temp2[df_temp2.columns[2]],palette=color_dict,style=df_temp2[df_temp2.columns[2]],marke
rs=markers_dict)
plt.title('Final Data Points')
plt.legend(loc=(1.05,0.75))
```

### Out[36]:

<matplotlib.legend.Legend at 0x162335714b0>



```
2 4 6 8 10

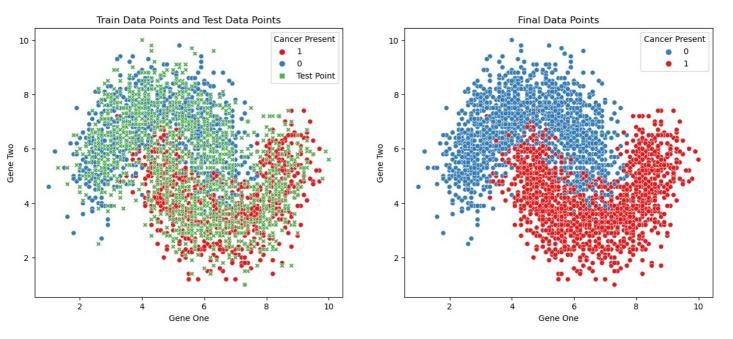
Gene One
```

### In [37]:

```
plt.figure(figsize=(15,6))
plt.subplot(1, 2, 1) # row 1, col 2 index 1
sns.scatterplot(x = df temp[df temp.columns[0]], y = df temp[df temp.columns[1]], hue=df temp[df temp.columns[1]]
emp[df_temp.columns[2]],palette=color_dict,style=df_temp[df_temp.columns[2]],markers=mar
plt.title('Train Data Points and Test Data Points')
# hue without palette : sns will provide default color for each group or class in df temp
 ['Cancer Present]
# huw with palette : sns will provide color we want for each group or class in df temp['C
ancer Present]
 # style without markers : sns will provide default shape for each group or class in df te
mp['Cancer Present]
# style with markers : sns will provide shape we want for each group or class in df temp[
 'Cancer Present]
plt.subplot(1, 2, 2) # index 2
sns.scatterplot(x = df temp2[df temp2.columns[0]], y = df temp2[df temp2.columns[1]], hue=df temp2[df temp2.columns[0]]  
equation (x = df temp2[df temp2.columns[0]], y = df temp2[df temp2.columns[1]], hue=df temp2[df temp2.columns[0]]  
equation (x = df temp2[df temp2.columns[0]], y = df temp2[df temp2.columns[1]], hue=df temp2[df temp2.columns[0]]  
equation (x = df temp2[df temp2.columns[0]], y = df temp2[df temp2.columns[0]], hue=df temp2[df temp2.columns[0]], 
f_temp2[df_temp2.columns[2]],palette=color_dict,style=df_temp2[df_temp2.columns[2]],marke
rs=markers dict)
plt.title('Final Data Points')
```

### Out[37]:

Text(0.5, 1.0, 'Final Data Points')



### In [41]:

right = 0

```
wrong = 0
for i in range(len(class name list)):
   if class_name_list[i] == df_test[df_test.columns[n-1]][i]:
       right = right + 1
   else:
       wrong = wrong + 1
print(len(df_test))
print(right, wrong)
1200
1189 11
In [39]:
accuarcy = (right * 100) / len(class_name_list)
print(accuarcy)
99.08333333333333
In [40]:
from sklearn.metrics import accuracy_score
accuracy score(df test[df test.columns[n-1]],class name list) * 100
Out[40]:
99.08333333333333
```