Assignment3

April 4, 2020

```
[]: import numpy as np
  import pandas as pd
  import matplotlib.pyplot as plt
  from sklearn.utils import shuffle
  from sklearn import svm
  import sys
# from plot import *
# set_things()
```

Q1. Plot the training data(Data1.xlsx) to get an idea of the data distribution. Plot the points with variable 1 on x-axis and variable 2 on y-axis. Now color the coordinates/points of class 0 with blue and class 1 with red. Report your visual observations.

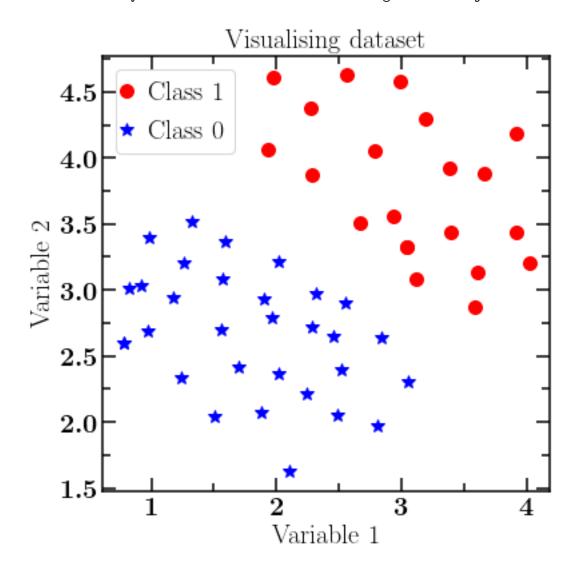
Loading dataset for Q1

```
[19]: df_train = pd.read_excel('Data1.xlsx')
```

Visualising the dataset

```
[20]: x1 = df_train['Variable 1'].values
     x2 = df_train['Variable 2'].values
     y = df_train['Class'].values
     for i in range(len(y)):
         if y[i] == -1:
             plt.plot(x1[i],x2[i],'-b*')
         if i==49:
             plt.plot(x1[i],x2[i],'-b*',label='Class 0')
         if y[i]==1:
             plt.plot(x1[i],x2[i],'-ro')
         if i==19:
             plt.plot(x1[i],x2[i],'-ro',label='Class 1')
     fontl = {'fontweight':1000,'fontsize':12}
     plt.legend(handlelength=0,frameon=True)
     plt.title("Visualising dataset")
     plt.xlabel("Variable 1")
     plt.ylabel("Variable 2")
     \# plt.savefig("Q1.png", dpi=300, facecolor='w', edgecolor='k', u)
      →bbox_inches='tight')
     plt.show()
```

findfont: Font family ['CMU Serif'] not found. Falling back to DejaVu Sans.



Q2. Apply SVM on training data (Data1.xlsx) to find decision boundary. Plot training data along with decision boundary

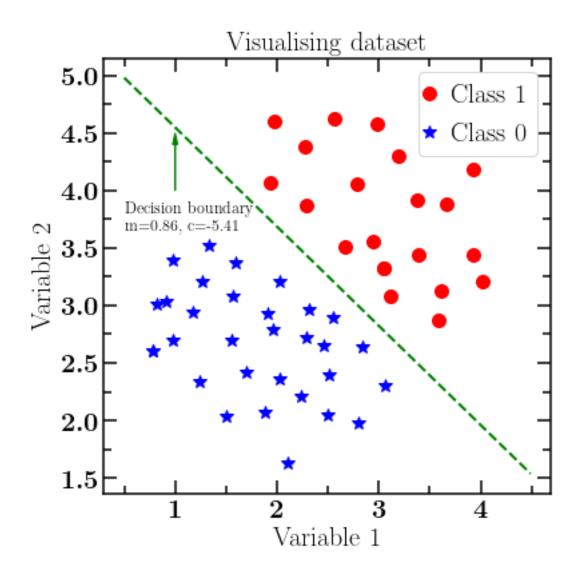
Solving for weights and bias

required to plot the decision boundary since the dataset is linearly separable

```
[21]: x1 = df_train['Variable 1'].values
    x2 = df_train['Variable 2'].values
    y = df_train['Class'].values
    for i in range(len(y)):
        if y[i]==-1:
            plt.plot(x1[i],x2[i],'-b*')
        if i==49:
```

```
plt.plot(x1[i],x2[i],'-b*',label='Class 0')
    if y[i]==1:
        plt.plot(x1[i],x2[i],'-ro')
    if i==19:
        plt.plot(x1[i],x2[i],'-ro',label='Class 1')
fontl = {'fontweight':1000,'fontsize':12}
plt.legend(handlelength=0,frameon=True)
plt.title("Visualising dataset")
plt.xlabel("Variable 1")
plt.ylabel("Variable 2")
X = df_train[df_train.columns[0:-1]]
y = df_train[df_train.columns[-1]]
C=1
clf = svm.SVC(kernel='linear',C=C).fit(X,y)
lin_svc=svm.LinearSVC(C=C).fit(X,y)
c=lin_svc.coef_
intercept=lin_svc.intercept_
xx = np.arange(0.5, 4.5, 0.01)
yy = -(c[0][0]/c[0][1])*xx - intercept/c[0][1]
plt.plot(xx,yy,'--g',label='Decision boundary')
# plt.legend()
ax = plt.axes()
ax.arrow(1.0,4.0,0,0.4,head_width=0.05, head_length=0.1, fc='g', ec='g')
plt.text(0.5,3.8,'Decision boundary',fontl)
plt.text(0.5,3.65, 'm=\%0.2f, c=\%0.2f'\%(c[0][0]/c[0][1], intercept/
 \rightarrowc[0][1]),fontsize=12)
# plt.savefig("Q2.png", dpi=300, facecolor='w', edgecolor='k',
→bbox_inches='tight')
plt.show()
```

//anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:30:
MatplotlibDeprecationWarning: Adding an axes using the same arguments as a previous axes currently reuses the earlier instance. In a future version, a new instance will always be created and returned. Meanwhile, this warning can be suppressed, and the future behavior ensured, by passing a unique label to each axes instance.

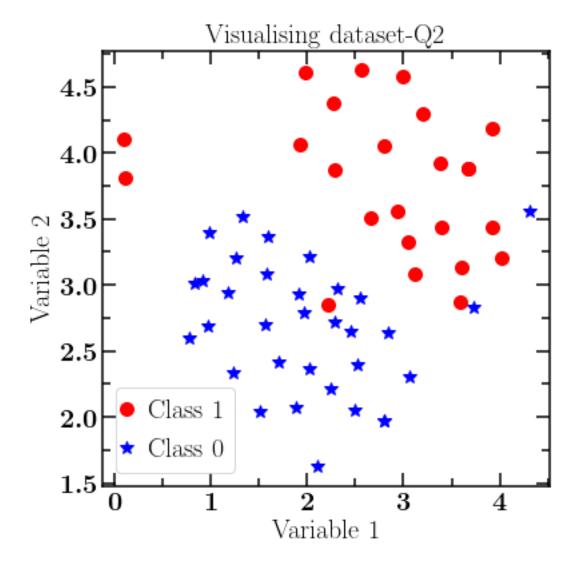


Visualising the dataset for Q2

```
[22]: df_2 = pd.read_excel('Data2.xlsx')

[23]: x1 = df_2['Variable 1'].values
    x2 = df_2['Variable 2'].values
    y = df_2['Class'].values
    for i in range(len(y)):
        if y[i]==-1:
            plt.plot(x1[i],x2[i],'-b*')
        if i==49:
            plt.plot(x1[i],x2[i],'-b*',label='Class 0')

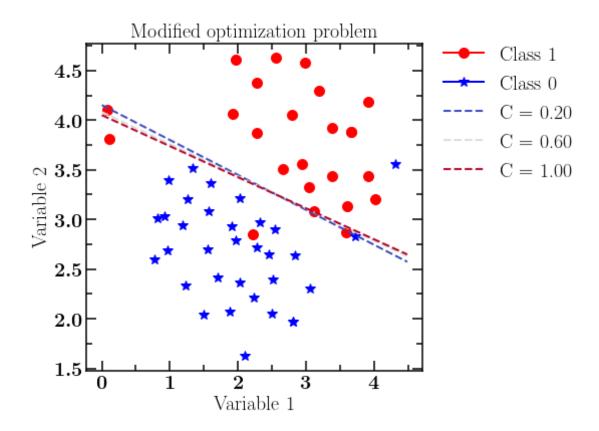
        if y[i]==1:
            plt.plot(x1[i],x2[i],'-ro')
        if i==19:
```

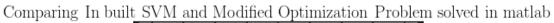


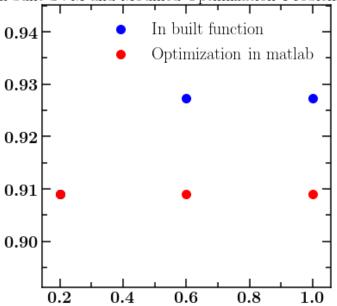
Q3. Now apply SVM with "modified optimization problem" on Data2.xlsx and try out different values of C and report your observations along with plots of the decision boundary. Modified Optimization Problem

```
[30]: #Plotting the data
     x1 = df 2['Variable 1'].values
     x2 = df 2['Variable 2'].values
     y = df_2['Class'].values
     for i in range(len(y)):
         if y[i] == -1:
             plt.plot(x1[i],x2[i],'-b*')
         if i==49:
             plt.plot(x1[i],x2[i],'-b*',label='Class 0')
         if y[i]==1:
             plt.plot(x1[i],x2[i],'-ro')
         if i==19:
             plt.plot(x1[i],x2[i],'-ro',label='Class 1')
     fontl = {'fontweight':1000,'fontsize':12}
     plt.legend(handlelength=0,frameon=True)
     plt.title("Modified optimization problem")
     plt.xlabel("Variable 1")
     plt.ylabel("Variable 2")
     #Modified Optimization Problem starts
     X = df_2[df_2.columns[0:-1]]
     y = df_2[df_2.columns[-1]]
     cc = plt.cm.coolwarm
     \# colors = cc(np.linspace(0, 1, len(C)))
     # C=[0.1,0.2,0.3,0.4,1,2,3,4,5]
     C = np.arange(0.1,1,0.1)
     C = np.array([0.2, 0.6, 1.0])
     scr = []
     colors = cc(np.linspace(0, 1, len(C)))
     for i in range(len(C)):
         lin_svc=svm.LinearSVC(C=C[i],max_iter=10000).fit(X,y)
         c=lin svc.coef
         intercept=lin_svc.intercept_
         xx = np.arange(0.0, 4.5, 0.01)
         yy = -(c[0][0]/c[0][1])*xx - intercept/c[0][1]
         plt.plot(xx,yy,color=colors[i],ls='--',label='C = %0.2f'%C[i])
         print(C[i],'%0.2f'%lin_svc.score(X,y))
         scr.append(lin_svc.score(X,y))
         # plt.legend()
           ax = plt.axes()
           ax.arrow(1.0,4.0,0,0.4,head\ width=0.05,\ head\ length=0.1,\ fc='q',\ ec='q')
     #
           plt.text(0.5,3.8, 'Decision boundary', fontl)
           plt.text(0.5,3.65, 'm=\%0.2f, c=\%0.2f'\%(c[0][0]/c[0][1], intercept/
      \hookrightarrow c[0][1]), fontsize=12)
```

0.2 0.91 0.6 0.93 1.0 0.93







[]: