

# HW6\_edgarsp2

November 11, 2019

## 1 STAT 542 / CS 598: Homework 6

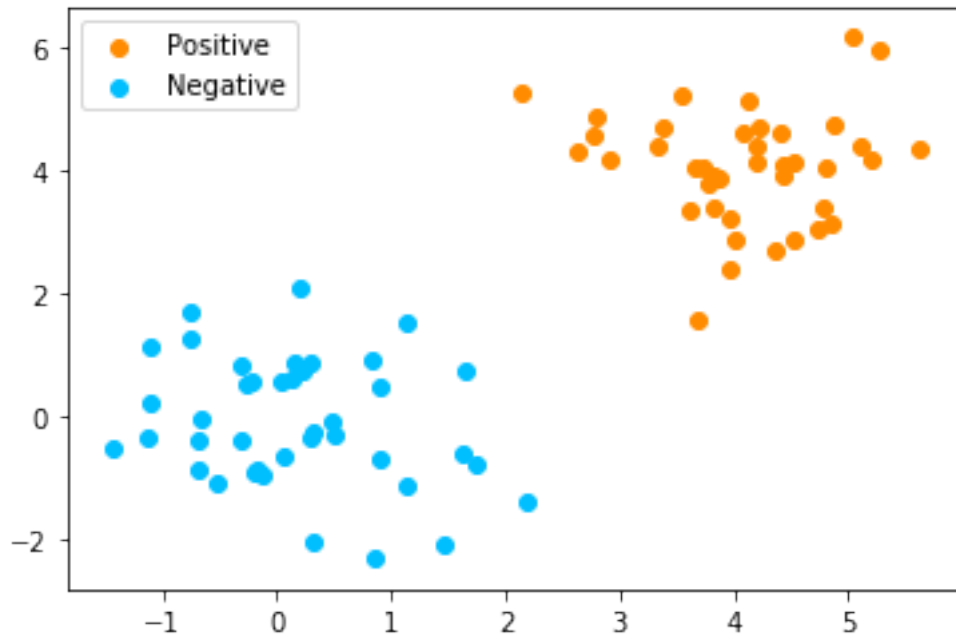
Fall 2019, by Edgar Pino

Due: Monday, Nov 11 by 11:59 PM Pacific Time

```
[2]: %matplotlib inline
[24]: import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from scipy.optimize import minimize
from cvxopt import matrix, solvers
import math
```

## 2 Question 1 [50 Points] Linearly Separable SVM using Quadratic Programming

```
[4]: np.random.seed(1)
[5]: COLOR_LABELS = ["darkorange", "deepskyblue"]
[6]: N = 40
P = 2
[7]: xpos = np.random.normal(0, 1, (N, P))
xneg = np.random.normal(4, 1, (N, P))
[8]: x = np.concatenate((xpos, xneg), axis=0)
y = np.concatenate((np.ones((N,)), -np.ones((N,))), axis=0)
[9]: def plot_data(x, y):
    unique = np.unique(y)
    for li in range(len(unique)):
        items = x[y == unique[li]]
        label = 'Positive' if unique[li] == -1 else 'Negative'
        plt.scatter(items[:, 0], items[:, 1], c = COLOR_LABELS[li], label=label)
    plt.legend()
[10]: plot_data(x, y)
```



### 2.0.1 Fit

```
[42]: def fit(x, y):
    NUM = x.shape[0]
    DIM = x.shape[1]
    K = y[:, None] * x
    K = np.dot(K, K.T)
    P = matrix(K)
    q = matrix(-np.ones((NUM, 1)))
    G = matrix(-np.eye(NUM))
    h = matrix(np.zeros(NUM))
    A = matrix(y.reshape(1, -1))
    b = matrix(np.zeros(1))

    return solvers.qp(P, q, G, h, A, b)
```

```
[43]: def plot_separator(ax, w, b):
    slope = -w[0] / w[1]
    intercept = -b / w[1]
    x = np.arange(0, 6)
    ax.plot(x, x * slope + intercept, 'k-')
    ax.plot(x, x * slope + intercept + 2.75, '-r')
    ax.plot(x, x * slope + intercept - 2.75, '-r')
```

```
[44]: def plot_data_with_labels(x, y, ax):
    unique = np.unique(y)
```

```

for li in range(len(unique)):
    x_sub = x[y == unique[li]]
    label = 'Positive' if unique[li] == 1 else 'Negative'
    ax.scatter(x_sub[:, 0], x_sub[:, 1], c = COLOR_LABELS[li], label=label)

```

```

[45]: sol = fit(x, y)
      alphas = np.array(sol['x'])

```

	pcost	dcost	gap	pres	dres
0:	-6.2205e+00	-1.0607e+01	2e+02	1e+01	2e+00
1:	-5.1204e+00	-1.6071e+00	3e+01	2e+00	2e-01
2:	-3.0840e-01	-5.4380e-01	5e-01	2e-02	2e-03
3:	-3.7365e-01	-4.2487e-01	9e-02	3e-03	3e-04
4:	-4.1115e-01	-4.1811e-01	9e-03	2e-04	2e-05
5:	-4.1661e-01	-4.1693e-01	3e-04	2e-06	2e-07
6:	-4.1685e-01	-4.1686e-01	6e-06	2e-08	2e-09
7:	-4.1685e-01	-4.1685e-01	6e-08	2e-10	2e-11

Optimal solution found.

```

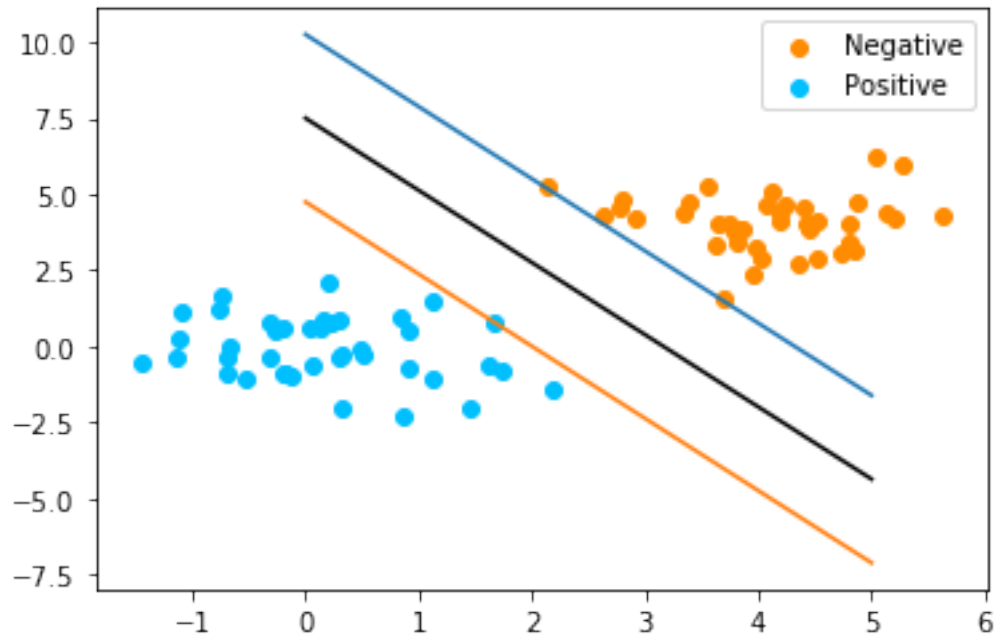
[46]: w = np.sum(alphas * y[:, None] * x, axis = 0)
      cond = (alphas > 1e-4).reshape(-1)
      b = y[cond] - np.dot(x[cond], w)
      bias = b[0]

      norm = np.linalg.norm(w)
      w, bias = w / norm, bias / norm

      fig, ax = plt.subplots()
      plot_separator(ax, w, bias)
      plot_data_with_labels(x, y, ax)

      plt.legend()
      plt.show()

```



### 3 Question 2 [25 Points] Linearly Non-seperable SVM using Penalized Loss

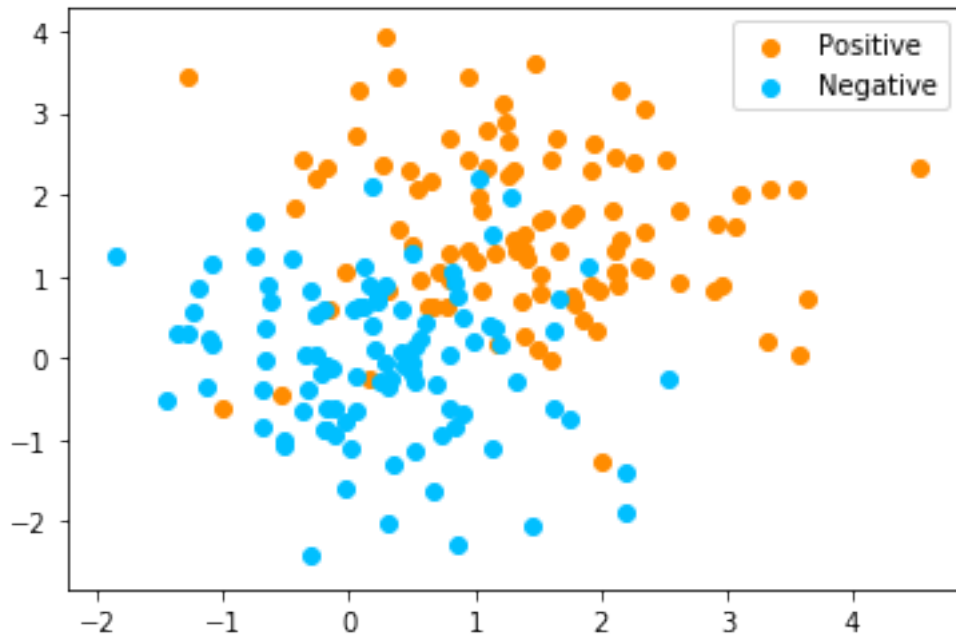
```
[51]: np.random.seed(1)
```

```
[52]: N = 100
      P = 2
```

```
[53]: xpos = np.random.normal(0, 1, (N,P))
      xneg = np.random.normal(1.5, 1, (N,P))
```

```
[54]: x = np.concatenate((xpos, xneg), axis=0)
      y = np.concatenate((np.ones((N,)), -np.ones((N,))), axis=0)
```

```
[55]: plot_data(x,y)
```



```
[56]: def loss(b, x, y):
      return np.sum(
          np.log(
              1+np.exp(
                  -y*
                  (b[0]+np.dot(x,np.array([b[1],b[2]])))
              ))
          )+0.5*(np.power(b[0],2)+np.power(b[1],2)+np.power(b[2],2))
```

```
[57]: results = minimize(loss, (1,1,1), args=(x,y))
```

```
[58]: intercept = -results['x'][0]/results['x'][2]
```

```
[59]: slope = -results['x'][1]/results['x'][2]
```

```
[60]: intercept
```

```
[60]: 1.3611124182617051
```

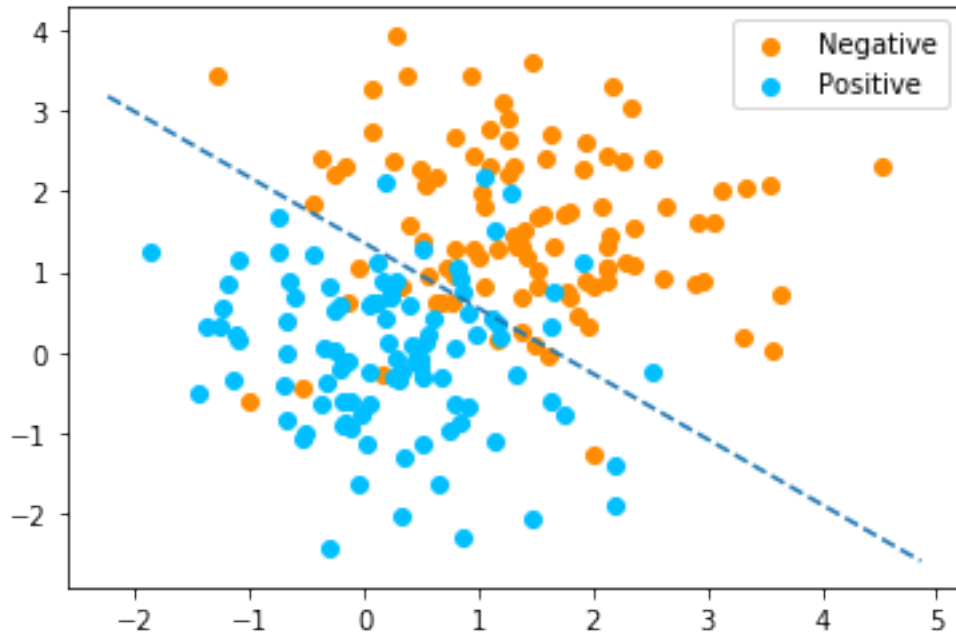
```
[61]: slope
```

```
[61]: -0.812670650533452
```

```
[62]: fig, ax = plt.subplots()
      plot_data_with_labels(x, y, ax)

      axes = plt.gca()
      x_vals = np.array(axes.get_xlim())
      y_vals = intercept + slope * x_vals
      ax.plot(x_vals, y_vals, '--')
```

```
plt.legend()
plt.show()
```



[ ]:

#### 4 Question 3 [25 Points] Nonlinear and Non-seperable SVM using Penalized Loss

```
[11]: from scipy.spatial.distance import pdist
      from rpy2.robjects.packages import STAP
      import rpy2.robjects.numpy2ri
      from rpy2.robjects.packages import importr
      rpy2.robjects.numpy2ri.activate()
```

```
[12]: import os
      os.environ['KMP_DUPLICATE_LIB_OK']='True'
```

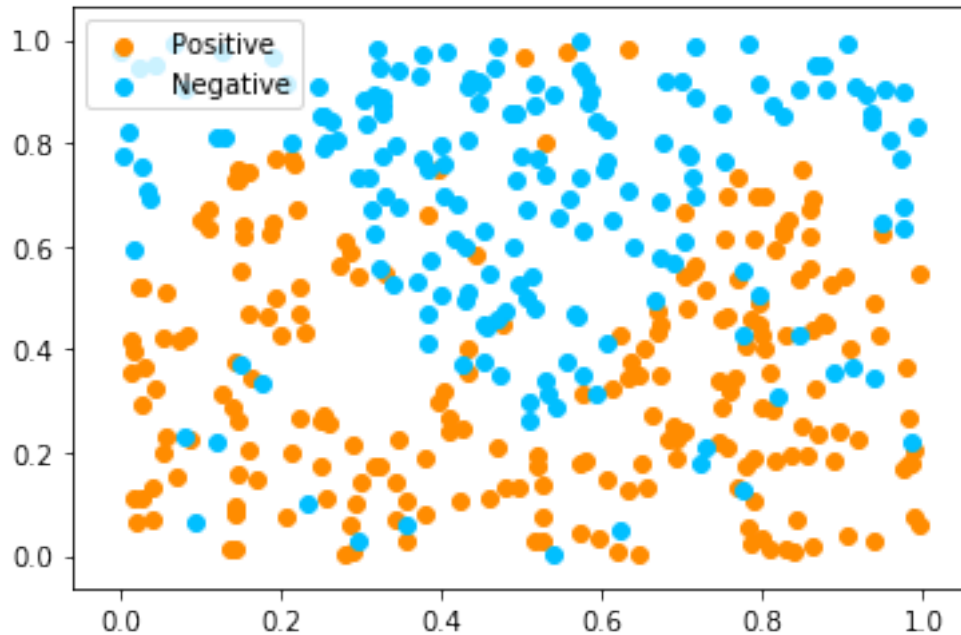
```
[13]: mfunc = 'get_k <- function(x){return(as.matrix(dist(x)^2)/0.25)}'
```

```
[14]: myasmatrix = STAP(mfunc, "myasmatrix")
```

```
[15]: N = 400
      P = 2
```

```
[16]: x = np.random.uniform(0,1,N*P).reshape(N,P)
```

```
[17]: side = (x[:, 1] > 0.5 + 0.3*np.sin(3*math.pi*x[:, 0]))
[18]: y = np.random.choice(np.array([1, -1]), N, True, np.array([0.9, 0.1]))*(side == 1) + np.random.choice(np.array([1, -1]), N, True, np.array([0.1, 0.9]))*(side == 0)
[19]: plot_data(x,y)
```



```
[20]: K = np.array(np.exp(-1 * np.array(myasmatrix.get_k(x))/ .25))
[76]: def my_k_loss(b, k, y):
        return np.sum(
            np.log(
                1+np.exp(
                    -y*np.dot(k,b)
                )
            )
        )+np.sum(0.05*b[k%b])
[77]: results = minimize(my_k_loss, np.ones(N), args=(K,y))
```

```
/anaconda3/lib/python3.7/site-packages/scipy/optimize/optimize.py:696:
RuntimeWarning: invalid value encountered in double_scalars
grad[k] = (f*((xk + d,) + args)) - f0) / d[k]
```

```
[79]: results['x']
```

[79]: array([1.15558281, 0.70875382, 0.63669815, 0.79667738, 0.85216392,  
0.83081225, 0.84137961, 0.72032158, 0.56282701, 0.73931824,  
0.75969523, 0.8130382 , 0.83313547, 0.59302644, 0.65992754,  
0.71233993, 0.83769635, 0.89331255, 0.85564678, 0.81355223,  
0.73191985, 0.70762637, 0.81347523, 0.71927339, 0.59213222,  
0.5980843 , 0.66520484, 0.63044827, 0.61124229, 0.76951721,  
0.65332069, 0.79852725, 0.85798987, 0.5617628 , 0.81162721,  
0.92121337, 0.78153629, 0.80277459, 0.65262116, 0.86834633,  
1.29268991, 0.73676209, 0.8385555 , 0.6523749 , 0.66611628,  
0.61270135, 0.58616137, 0.85472188, 0.64655303, 0.67138266,  
0.56904923, 0.83162056, 0.74578153, 1.04213048, 0.67870159,  
0.71208931, 0.75320888, 0.79521885, 0.67741315, 0.702786 ,  
0.82368113, 0.62970758, 0.86374028, 1.31426082, 0.73519383,  
0.67212268, 0.68532017, 0.74072777, 0.61720687, 0.57829385,  
0.57903322, 0.86683974, 0.62605405, 0.66027202, 0.75668138,  
0.73713743, 0.61690083, 0.75103914, 0.76783306, 0.70103815,  
0.72543048, 0.57548963, 0.73471236, 0.75361103, 0.61849657,  
0.57998583, 0.7098075 , 0.78479017, 0.69523624, 0.81972859,  
0.78737078, 0.7097401 , 0.82264276, 0.78431503, 0.84866025,  
0.62163203, 0.68221662, 0.72755932, 0.7174603 , 0.55423288,  
0.59159912, 0.6732506 , 0.71563074, 0.77201604, 0.63158439,  
0.80397743, 0.78864517, 0.68010466, 0.68678498, 0.95923493,  
0.73145427, 0.61473248, 0.69319686, 1.00367414, 0.86587116,  
0.84108271, 0.65884096, 0.70838442, 0.88463317, 0.74797158,  
0.62803947, 0.62306683, 0.80346811, 0.63895012, 0.57535216,  
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0.83758071, 0.72854795, 0.81911373, 0.7659454 , 0.75281677,  
0.7052176 , 0.591633 , 0.8419964 , 0.61223551, 0.84516312,  
0.83798488, 0.64583305, 0.67229606, 0.61708582, 0.61134366,  
0.59889236, 0.55978286, 0.60244959, 0.69542959, 0.62263562,  
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0.84571263, 0.69726276, 0.82928048, 0.68913852, 0.66352207,  
0.69691054, 0.56616597, 0.68606745, 0.55263171, 0.57032293,  
0.56436256, 0.77099601, 0.76684714, 0.56692635, 0.62518303,  
0.65164109, 0.79756595, 0.79188739, 0.68164673, 0.83951596,  
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0.63905352, 0.66237537, 0.77970275, 0.71625123, 0.5831348 ,  
0.78241923, 0.68963192, 0.76361527, 0.66166683, 0.6678408 ,  
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0.6629127 , 0.77235963, 1.00732243, 0.62556691, 0.66533524,  
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 0.69641678, 0.82828381, 0.64350653, 0.83562934, 1.03645137,  
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 0.67155376, 0.75425709, 0.72849339, 0.65169169, 0.83289332,  
 0.66268288, 0.70133248, 0.64136686, 0.82111618, 0.57429242,  
 0.61380465, 0.60908131, 1.0635679 , 0.61026012, 0.85747152,  
 0.7389343 , 0.7183351 , 0.65547081, 0.74066176, 0.76170961,  
 0.95977351, 0.88675008, 0.69624981, 0.63561203, 0.75521627,  
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 0.72489097, 0.77564554, 0.58994977, 0.61509671, 0.7449496 ])

[ ]: