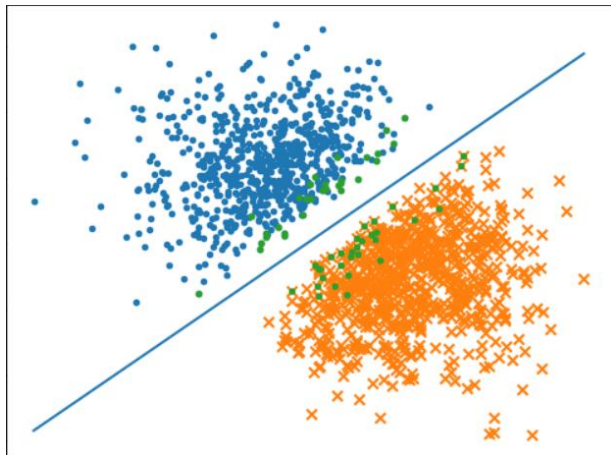
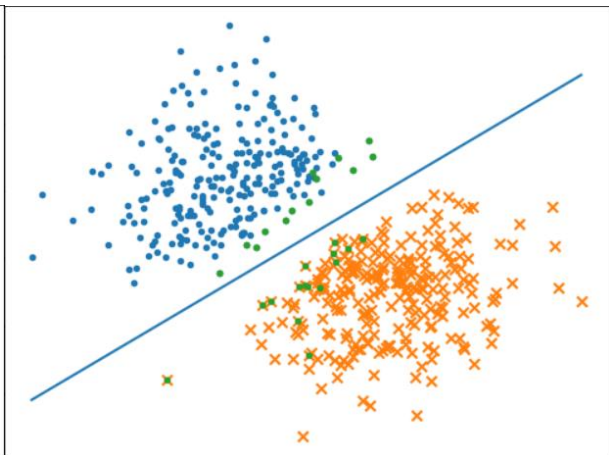
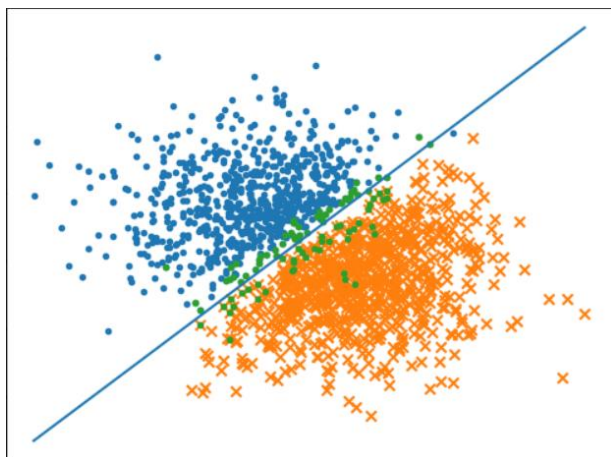
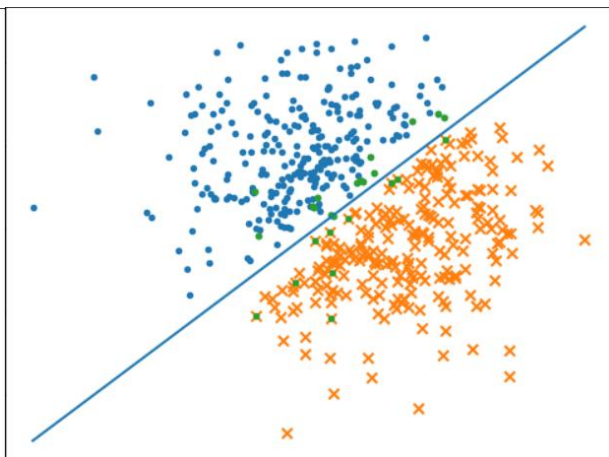


# 山东大学 计算机科学与技术 学院

## 机器学习 课程实验报告

学号：201618130133		姓名：代荣森	班级：2016. 4
实验题目：SVM			
实验学时：4		实验日期：2018. 11. 30	
实验目的： 学习使用支持向量机进行线性和非线性分类			
硬件环境： i5-6200U 8G RAM HD Graphics520			
软件环境： Visual Studio Code + Python			
实验步骤与内容： 本次三个实验均使用 SMO 代替标准 QP 求解器。			
实验一			
1、用训练数据绘制 SVM 的决策边界。			
训练集 1		测试集 1	
			
准确率：1.0			
训练集 2			
			
准确率：0.996			
绿色的点为支持向量			
出错在第 109 和 278 行			

2、尝试正则化项  $C$  的二元数值，并报告您的观察结果。

改变  $C$  的值对结果影响很小。

### 实验二、手写数字识别

由于原数据集过大，故随机抽取 200 个数据进行训练。

用训练集直接测试得到正确率为

训练值数量: 200

准确率: 1.0

用测试集进行测试得到正确率为

训练值数量: 200

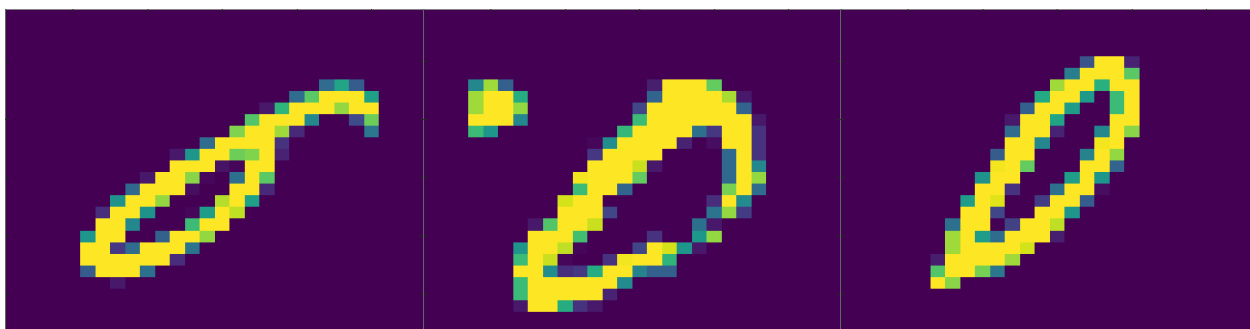
准确率: 0.9985815602836879

用测试集训练测试得到正确率为

训练值数量: 2115

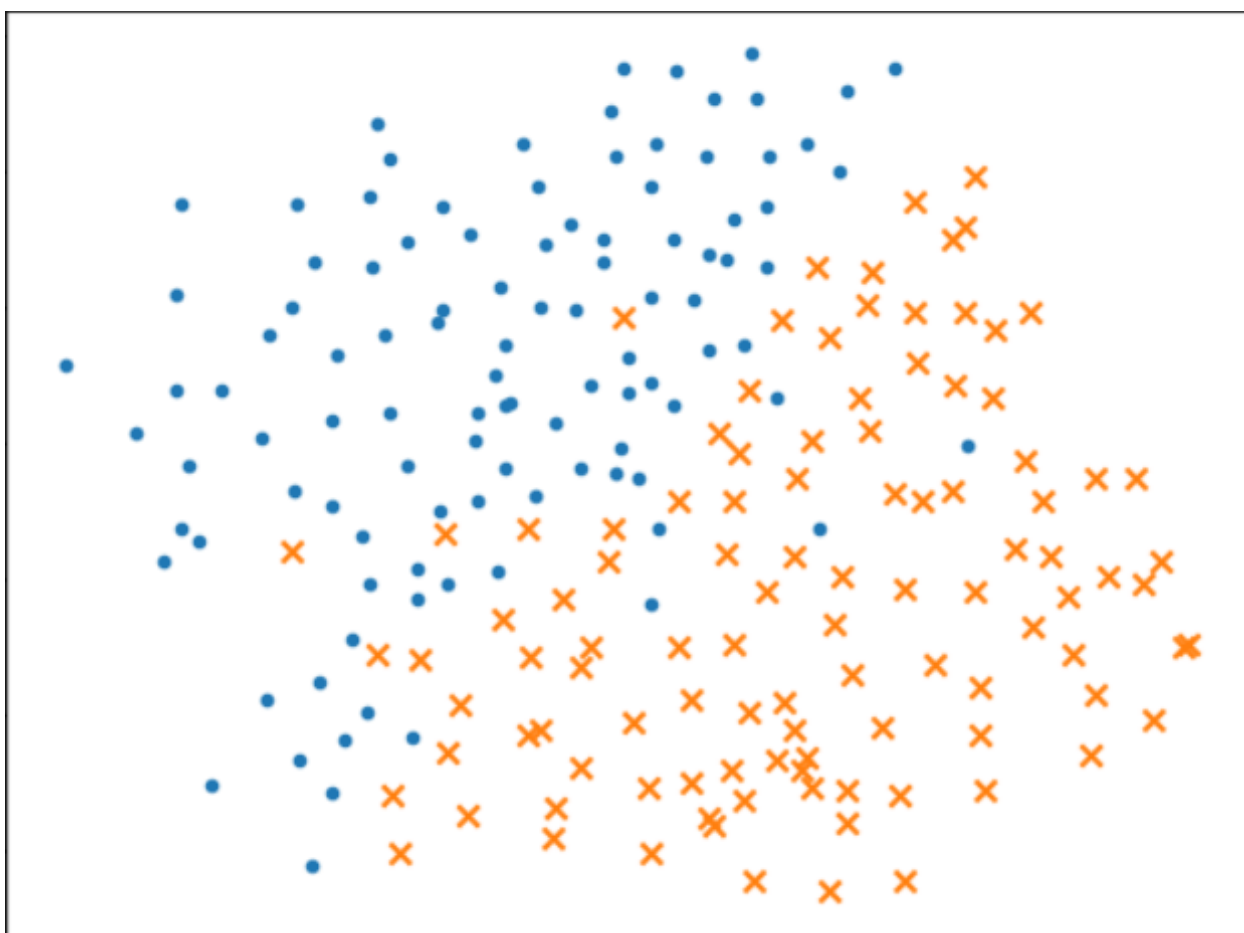
准确率: 1.0

得到 3 个错误图像



### 实验三、非线性 SVM

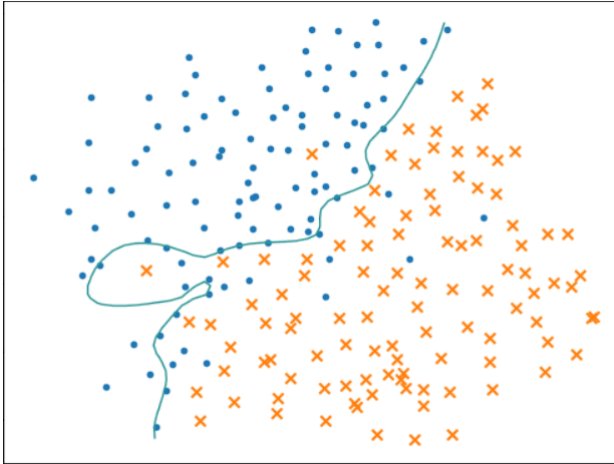
绘制散点图，发现其线性不可分



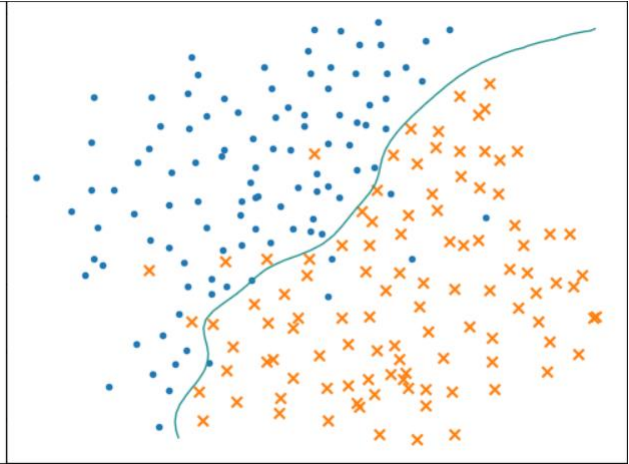
### 使用加入高斯内核后的 SMO

```
for xi in range(length):
    for xj in range(length):
        zij[xi,xj] = np.exp(-gama * np.sum((x[xi,:] - x[xj,:])**2))
```

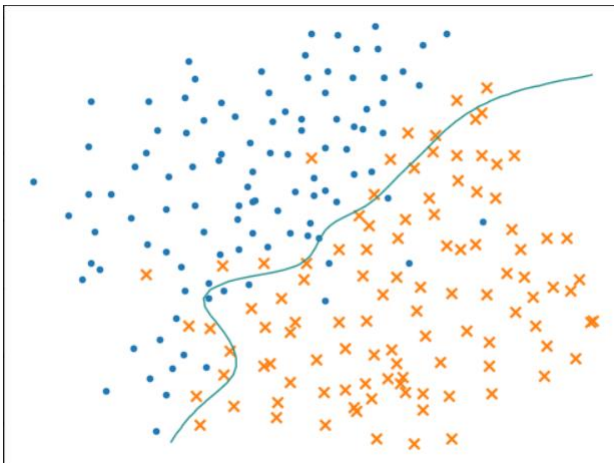
当  $C = 10$ ,  $\gamma = 100$  时:



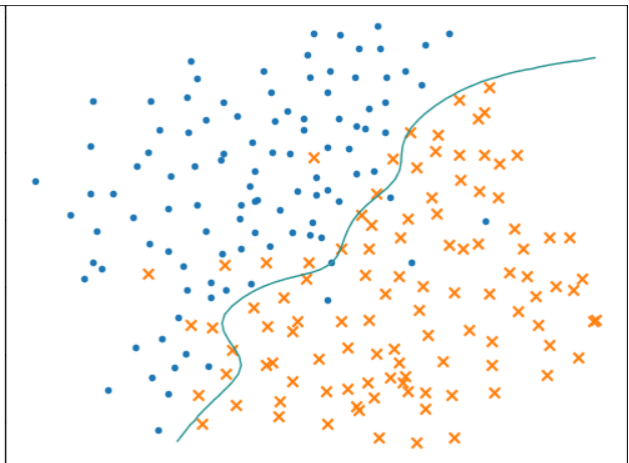
当  $C = 10$ ,  $\gamma = 1000$  时:



当  $C = 1$ ,  $\gamma = 100$  时:



当  $C = 1$ ,  $\gamma = 1000$  时:



### 结论分析与体会:

最近在学习 python, 所以使用 python 来写了这个实验, 由于一些规则用法不一样, 所以有些地方需要特别注意

之前用 libSVM 库直接可以得出训练测试结果, 但由于完全交给库来做, 能学到的只有对训练和测试集进行字符串处理, 最后又决定重新用 SMO 算法来完成实验

实验难度主要在对于 SMO 和核函数的理解上, 以及对于算法算出的  $\alpha$  和支持向量如何画图。写完之后对于 SMO 还是有很多地方不是很清楚, 还需多多体会。

### 附录: 程序源代码

SMO.py

```
import numpy as np
def SMO(x, y, C):
    # x = np.matrix(x)
    length = len(y)
    alpha = np.zeros(length)
    b = 0
```

```

tol = 1e-7
while 1:
    num_changed_alphas = 0
    xt = np.transpose(x)
    xt = np.matrix(xt)
    #计算内积
    zij = x * xt
    for i in range(length):
        Ei = alpha * y * zij[:,i] + b - y[i]
        #不满足 kkt 条件
        if (y[i] * Ei < -tol and alpha[i] < C) or (y[i] * Ei > tol and
alpha[i] > 0):
            for j in range(0,i):
                Ej = alpha * y * zij[:,j] + b - y[j]
                alpha_old_i = alpha[i]
                alpha_old_j = alpha[j]
                if y[i] == y[j]:
                    L = np.maximum(0, alpha[i] + alpha[j] - C)
                    H = np.minimum(C, alpha[j] + alpha[i])
                else:
                    L = np.maximum(0, alpha[j] - alpha[i])
                    H = np.minimum(C, C + alpha[j] - alpha[i])
                if L == H:
                    continue
                yita = zij[i,j] * 2 - zij[i,i] - zij[j,j]
                if yita >= 0:
                    continue
                alpha[j] = alpha[j] - y[j] * (Ei - Ej) / yita
                if alpha[j] > H:
                    alpha[j] = H
                elif alpha[j] < L:
                    alpha[j] = L
                if np.linalg.norm(alpha[j] - alpha_old_j) < tol:
                    continue
                alpha[i] = alpha[i] + y[i] * y[j] * (alpha_old_j - alpha[j])
                b1 = b - Ei - y[i] * (alpha[i] - alpha_old_i) * zij[i,i] - y[j]
* (alpha[j]-alpha_old_j) * zij[i,j]
                b2 = b - Ej - y[i] * (alpha[i] - alpha_old_i) * zij[i,j] - y[j]
* (alpha[j]-alpha_old_j) * zij[j,j]
                if alpha[i] < C and alpha[i] > 0:
                    b = b1
                elif alpha[j] < C and alpha[j] > 0:
                    b = b2
                else:
                    b = (b1 + b2) / 2
            #重新计算 Ei,如果还不满足 kkt 条件,继续,如果满足则寻找下一个 i

```

```

        Ei = alpha * y * zij[:,i] + b - y[i]
        if (y[i] * Ei < -tol and alpha[i] < C) or (y[i] * Ei > tol and
alpha[i] > 0):
            num_changed_alphas = num_changed_alphas + 1
        else:
            break
    #去除 alpha==0 和 alpha==c
    x1 = []
    y1 = []
    alpha1 = []
    for k in range(len(alpha)):
        if alpha[k] != 0 and alpha[k] != C:
            # la[k] = 1
            x1.append(x[k])
            y1.append(y[k])
            alpha1.append(alpha[k])

    x = x1
    y = np.array(y1)
    alpha = np.array(alpha1)
    length = len(y)
    if num_changed_alphas == 0:
        break
    return alpha,x,y,b #x:支持向量, y:label

```

Gaussiansmo.py

```

import numpy as np
def Gaussiansmo(x, y, C, gama):
    # x = np.matrix(x)
    length = len(y)
    alpha = np.zeros(length)
    b = 0
    tol = 1e-7
    while 1:
        num_changed_alphas = 0
        zij = np.zeros((length,length))
        zij = np.matrix(zij)
        for xi in range(length):
            for xj in range(length):
                zij[xi,xj] = np.exp(-gama * np.sum((x[xi,:] - x[xj,:])**2))

        for i in range(length):
            Ei = alpha * y * zij[:,i] + b - y[i]
            #不满足 kkt 条件
            if (y[i] * Ei < - tol and alpha[i] < C) or (y[i] * Ei > tol and
alpha[i] > 0):
                for j in range(0,length,2):

```

```

Ej = alpha * y * zij[:,j] + b - y[j]
alpha_old_i = alpha[i]
alpha_old_j = alpha[j]
if y[i] == y[j]:
    L = np.maximum(0, alpha[i] + alpha[j] - C)
    H = np.minimum(C, alpha[j] + alpha[i])
else:
    L = np.maximum(0, alpha[j] - alpha[i])
    H = np.minimum(C, C + alpha[j] - alpha[i])
if L == H:
    continue
yita = zij[i,j] * 2 - zij[i,i] - zij[j,j]
if yita >= 0:
    continue
alpha[j] = alpha[j] - y[j] * (Ei - Ej) / yita
if alpha[j] > H:
    alpha[j] = H
elif alpha[j] < L:
    alpha[j] = L
if np.linalg.norm(alpha[j] - alpha_old_j) < tol:
    continue
alpha[i] = alpha[i] + y[i] * y[j] * (alpha_old_j - alpha[j])
b1 = b - Ei - y[i] * (alpha[i] - alpha_old_i) * zij[i,i] - y[j]
* (alpha[j]-alpha_old_j) * zij[i,j]
b2 = b - Ej - y[i] * (alpha[i] - alpha_old_i) * zij[i,j] - y[j]
* (alpha[j]-alpha_old_j) * zij[j,j]
if alpha[i] < C and alpha[i] > 0:
    b = b1
elif alpha[j] < C and alpha[j] > 0:
    b = b2
else:
    b = (b1 + b2) / 2
#重新计算 Ei,如果还不满足 kkt 条件,继续,如果满足则寻找下一个 i
Ei = alpha * y * zij[:,i] + b - y[i]
# if (y[i] * Ei <- tol and alpha[i] < C) or (y[i] * Ei > tol
and alpha[i] > 0):
    # num_changed_alphas = num_changed_alphas + 1
# else:
# break
num_changed_alphas = num_changed_alphas + 1
#去除 alpha==0 和 alpha==c
# la = alpha
x1 = []
y1 = []
alpha1 = []
for k in range(len(alpha)):

```

```
        if alpha[k] != 0 and alpha[k] != C:
            # la[k] = 1
            x1.append(x[k])
            y1.append(y[k])
            alpha1.append(alpha[k])

    x = np.array(x1)
    y = np.array(y1)
    alpha = np.array(alpha1)
    length = len(y)
    if num_changed_alphas == 0:
        break
    return alpha, x, y, b #x:支持向量, y:label
```