# Homework1

January 14, 2021

#### 0.0.1 Qinhua Sun

# 1 Problem 1: Python & Data Exploration

```
[6]: import numpy as np import matplotlib.pyplot as plt
```

```
[7]: iris = np.genfromtxt("data/iris.txt",delimiter=None) # load the text file
Y = iris[:,-1] # target value is the last column
X = iris[:,0:-1] # features are the other columns
```

Question 1

```
[8]: print(X.shape)
```

(148, 4)

There are 148 data points and 4 features.

#### 1.1 Question 2

```
[9]: # slicing the data into 3 groups
X1 = X[Y==0]
X2 = X[Y==1]
X3 = X[Y==2]

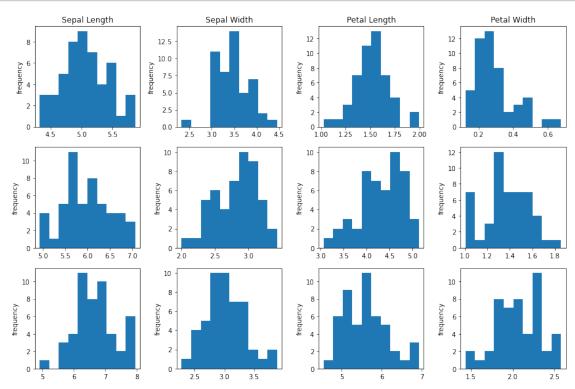
# # plot the data
fig, ax = plt.subplots(3, 4, tight_layout=True, figsize=(12,8))
features = ['Sepal Length', 'Sepal Width', 'Petal Length', 'Petal Width']

for i in range(4):
    ax[0,i].hist(X1[:,i])
    ax[0,i].set_title(features[i])

for i in range(4):
    ax[1,i].hist(X2[:,i])

for i in range(4):
    ax[2,i].hist(X3[:,i])
```

```
for ax in ax.flat:
    ax.set(ylabel='frequency')
```



### 1.2 Question 3

```
[10]: print('Mean:',np.mean(X, axis=0))
    print('Median:',np.median(X, axis=0))
    print('Variance', np.var(X, axis=0))
    print('SD', np.std(X, axis=0))
```

Mean: [5.90010376 3.09893092 3.81955484 1.25255548]
Median: [5.84664255 3.05980605 4.3998377 1.361768 ]
Variance [0.694559 0.19035057 3.07671634 0.57573564]
SD [0.83340207 0.43629184 1.75405711 0.75877246]

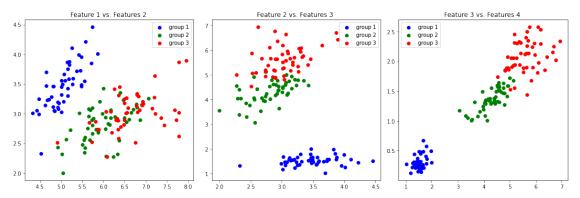
#### 1.3 Question 4

```
[11]: import more_itertools
fig, ax = plt.subplots(1, 3, tight_layout=True, figsize=(15,5))
colors = ["b", "g", "r"]

for i in range(3):
```

```
ax[0].scatter(X[Y==i][:,0],X[Y==i][:,1], color = colors[i], label = 'group_

$\d' \% \text{ int(i+1)}\\
ax[0].set_title('Feature 1 vs. Features 2 ')
ax[0].legend()
ax[1].scatter(X[Y==i][:,1],X[Y==i][:,2], color = colors[i], label = 'group_
$\d' \% \text{ int(i+1)}\\
ax[1].set_title('Feature 2 vs. Features 3 ')
ax[1].legend()
ax[2].scatter(X[Y==i][:,2],X[Y==i][:,3], color = colors[i], label = 'group_
$\d' \% \text{ int(i+1)}\\
ax[2].set_title('Feature 3 vs. Features 4 ')
ax[2].legend()
```



## 2 Problem 2

#### 2.1 Question 1

When the matrix is square and the determinant of the matrix is not zero.

### 2.2 Question 2 - 5

#2 Det (A), Det (B)

A = 
$$\begin{bmatrix} 1 & 2 & 3 \\ 2 & -1 & 1 \\ 1 & 3 & 2 \end{bmatrix}$$

B =  $\begin{bmatrix} 0 & -8 & -2 \\ 1 & -4 & -2 \\ -4 & 4 & 1 \end{bmatrix}$ 
 $|A| = | \times (|A| \times 2 - 3 \times 1) - 2 \times (2 \times 2 - 1 \times 1) + 3 \times (2 \times 3 - 1 \times (41))$ 
 $= | (-5) - 2 \times 3 + 3 \times 7 + 2 + 3 \times 7 + 3$ 

Mathix of numbers of B.

$$\begin{bmatrix}
(4) \times 1 - 4 \times (-2) & | \times 1 - (-1) \times (-2) & | \times 4 - (-4) \times (-4) \\
(-8) \times (-1) - 4 \times (-2) & | \times 1 - (-4) \times (-2) & | \times 4 - (-4) \times (-8) \\
(-6) \times (-2) - (4) \times (-2) & | \times (-2) - | \times 1 \times 2) & | \times (-4) \times (-4) & | \times (-4) \\
= \begin{bmatrix}
4 & -7 - 8 \\
-8 & -32 \\
8 & 2 & 8
\end{bmatrix} = \begin{bmatrix}
4 & 7 - 8 \\
-8 & -32 \\
8 & -2 & 8
\end{bmatrix} = \begin{bmatrix}
4 & 7 - 8 \\
-8 & -32 \\
8 & -2 & 8
\end{bmatrix}$$

$$B^{-1} = \frac{1}{-32} \begin{bmatrix}
4 & 9 & 8 \\
7 & -8 & -32 \\
8 & -2 & 8
\end{bmatrix} = \begin{bmatrix}
-\frac{1}{3} & 0 - \frac{1}{4} \\
-\frac{1}{3} & 4 - \frac{1}{6}
\end{bmatrix}$$

$$(A^{T})^{-1} = \begin{bmatrix}
A^{-1})^{T} = \begin{bmatrix}
-6 & -\frac{1}{4} & -\frac{1}{16} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{4}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1})^{T} = \begin{bmatrix}
-\frac{1}{8} & -\frac{1}{3} & \frac{1}{4} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{4}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{4} & -\frac{1}{16} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{4}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{4} & -\frac{1}{16} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{4}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{4} & -\frac{1}{16} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{4} & -\frac{1}{16} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{4} & -\frac{1}{16} \\
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\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{4} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{-1} & -\frac{1}{16} & -\frac{1}{16} \\
-\frac{1}{16} & -\frac{1}{16} & -\frac{1}{16}
\end{bmatrix}$$

$$(B^{T})^{-1} = \begin{bmatrix}
B^{1$$

## 3 Probelm 3

```
[52]: import mltools as ml
    np.random.seed(0)

iris = np.genfromtxt("data/iris.txt", delimiter=None)
    Y = iris[:,-1]
    X = iris[:,0:2]

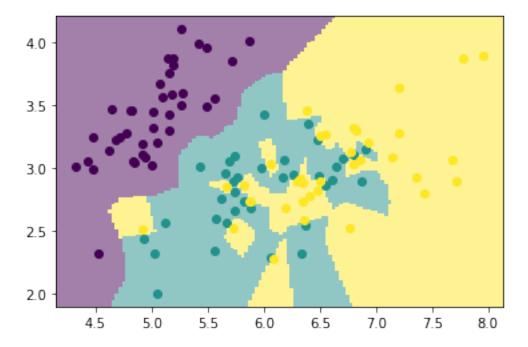
# shuffle the data
    X,Y = ml.shuffleData(X,Y)

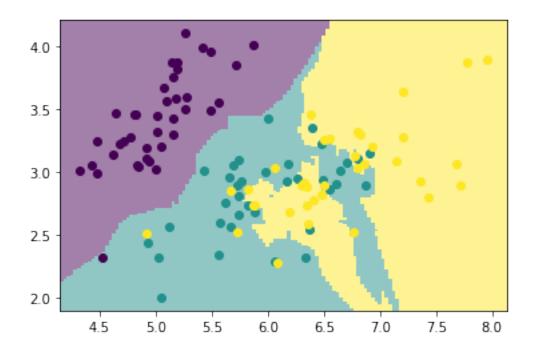
# split the data into 80/20 train/validation set
    Xtr,Xva,Ytr,Yva = ml.splitData(X,Y, 0.75);
```

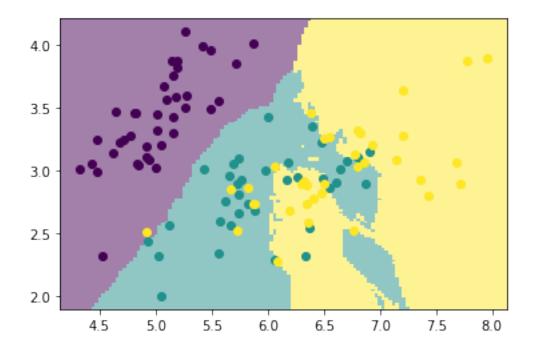
## 3.1 Question 1

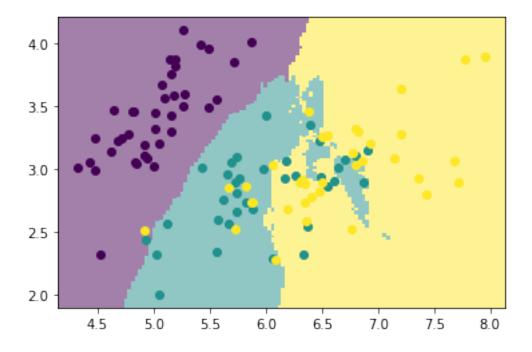
```
[67]: k= [1,5,10,15,50]

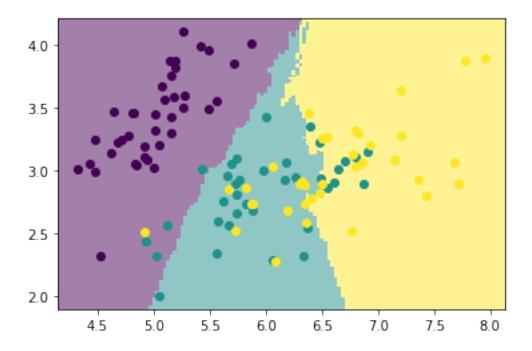
for i,k in enumerate(k):
    knn = ml.knn.knnClassify()
    knn.train(Xtr, Ytr, k)
    YvaHat = knn.predict(Xva)
    ml.plotClassify2D(knn, Xtr, Ytr)
```











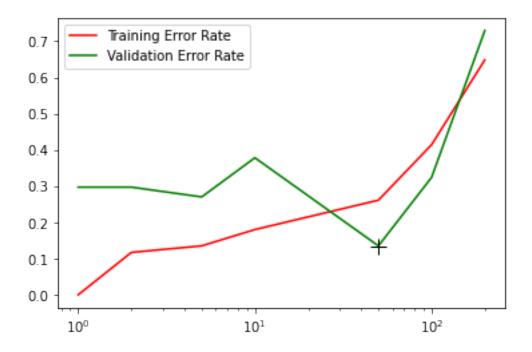
As we increase the value of K, wea're averaging more neighbors and therefore will have smoother decision boundary. However, the error will also be increasing.

#### 3.2 Question 2

```
[128]: K = [1,2,5,10,50,100,200];
       errTrain =[]
       errVa = []
       for i,k in enumerate(K):
           learner = ml.knn.knnClassify()
           k = K[i]
           learner.train(Xtr, Ytr, k)
           Yhat = learner.predict(Xtr)
           errTrain.append(np.mean(Yhat != Ytr))
           YvaHat = learner.predict(Xva)
           errVa.append(np.mean(YvaHat != Yva))
       print(errTrain)
       print(errVa)
       plt.semilogx(K, errTrain, "r", label = "Training Error Rate")
       plt.semilogx(K, errVa, "g", label = "Validation Error Rate" )
       plt.legend()
       plt.semilogx(K[np.argmin(errVa)],min(errVa),'k+',markersize=12)
       print('\n')
       print(K[np.argmin(errVa)], 'is when validation error is the lowest')
```

[0.0, 0.11711711711711711, 0.13513513513513514, 0.18018018018018017, 0.26126126126126126, 0.4144144144144144, 0.6486486486486487] [0.2972972972973, 0.2972972972973, 0.2702702702702703, 0.3783783783783784, 0.13513513513513514, 0.32432432432432434, 0.7297297297297297]

50 is when validation error is the lowest



I will choose k=50 as it has the lowest validation error rate.

### 3.3 Question 3

```
[129]: import mltools as ml
       np.random.seed(0)
       iris = np.genfromtxt("data/iris.txt", delimiter=None)
       Y = iris[:,-1]
       X = iris[:,0:-1]
       # shuffle the data
       X,Y = ml.shuffleData(X,Y)
       # split the data into 80/20 train/validation set
       Xtr,Xva,Ytr,Yva = ml.splitData(X,Y, 0.75);
       K = [1,2,5,10,50,100,200];
       errTrain =[]
       errVa = []
       for i,k in enumerate(K):
           learner = ml.knn.knnClassify()
           k = K[i]
           learner.train(Xtr, Ytr, k)
```

```
Yhat = learner.predict(Xtr)
  errTrain.append(np.mean(Yhat != Ytr))
YvaHat = learner.predict(Xva)
  errVa.append(np.mean(YvaHat != Yva))

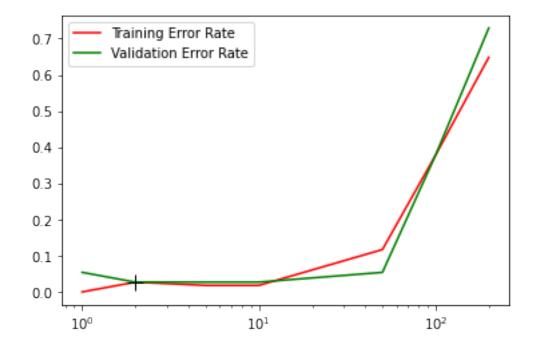
print(errTrain)
print(errVa)

plt.semilogx(K, errTrain, "r", label = "Training Error Rate")
plt.semilogx(K, errVa, "g", label = "Validation Error Rate")
plt.legend()

plt.semilogx(K[np.argmin(errVa)],min(errVa),'k+',markersize=12)
print('\n')
print(K[np.argmin(errVa)],'is when validation error is the lowest')
```

[0.0, 0.02702702702702703, 0.018018018018018018, 0.018018018018018018, 0.117117117117117, 0.3783783783783783784, 0.6486486486486487] [0.05405405405405406, 0.02702702702702703, 0.02702702702702703, 0.02702702702703, 0.05405405405405406, 0.3783783783783784, 0.7297297297297]

#### 2 is when validation error is the lowest



The plot generated by using all four features looks very different. Now I would recommand using

## 4 Problem 4

### 4.1 Question 1

Class probability:

$$p(y = +1) = \frac{4}{10} = \frac{2}{5}$$
$$p(y = -1) = \frac{6}{10} = \frac{3}{5}$$

Individual feature probability:

$$P(X_{1} = 0|y = +1) = \frac{1}{4}$$

$$P(X_{1} = 1|y = +1) = \frac{3}{4}$$

$$P(X_{1} = 0|y = -1) = \frac{1}{2}$$

$$P(X_{1} = 1|y = -1) = \frac{1}{2}$$

$$P(X_{2} = 0|y = +1) = 1$$

$$P(X_{2} = 1|y = +1) = 0$$

$$P(X_{2} = 0|y = -1) = \frac{1}{6}$$

$$P(X_{2} = 1|y = -1) = \frac{5}{6}$$

$$P(X_{3} = 0|y = +1) = \frac{1}{4}$$

$$P(X_{3} = 1|y = +1) = \frac{3}{4}$$

$$P(X_{3} = 1|y = +1) = \frac{3}{4}$$

$$P(X_{3} = 1|y = -1) = \frac{1}{3}$$

$$P(X_{4} = 0|y = +1) = \frac{1}{2}$$

$$P(X_{4} = 0|y = -1) = \frac{1}{6}$$

$$P(X_{4} = 1|y = -1) = \frac{5}{6}$$

$$P(X_5 = 0|y = +1) = \frac{3}{4}$$

$$P(X_5 = 1|y = +1) = \frac{1}{4}$$

$$P(X_5 = 0|y = -1) = \frac{2}{3}$$

$$P(X_5 = 1|y = -1) = \frac{1}{3}$$

### 4.2 Question 2

$$p(y=1|x=(00000)=p(y=+1)\times\frac{1}{4}\times1\times\frac{1}{4}\times\frac{1}{2}\times\frac{3}{4}=0.0093$$
 
$$p(y=-1|x=(00000)=p(y=-1)\times\frac{1}{2}\times1\times\frac{1}{6}\times\frac{1}{3}\times\frac{1}{6}\times\frac{2}{3}=0.0019$$

 $x = (0\ 0\ 0\ 0\ 0)$  would be predicted for as +1, and  $x = (1\ 1\ 0\ 1\ 0)$  would be -1.

#### 4.3 Question 3

$$p(y = +1|x = (11010) = \frac{p(x = (11010)|y = +1)) \times p(y = +1)}{p(x = (11010))}$$

since

$$p(x = (11010)|y = +1) = \frac{3}{4} \times 0 \times \dots$$

The posterior probability is virtually 0.

#### 4.4 The remaining questions will be submitted through next submission.

# Statement of Collaboration

This is the Statement of Collaboration, meaning that I have followed the academic honesty guidelines and did not discussed this assignment with anyone.

Qinhua Sun