CS184A Homework 1 Report

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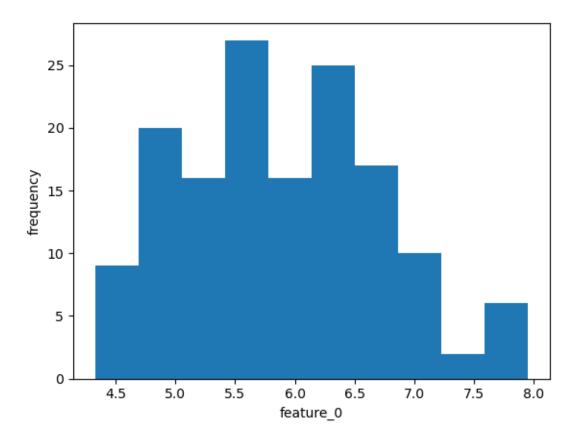
Problem 1: Python & Data Exploration

```
[]: import numpy as np
  import matplotlib.pyplot as plt
  np.random.seed(42)

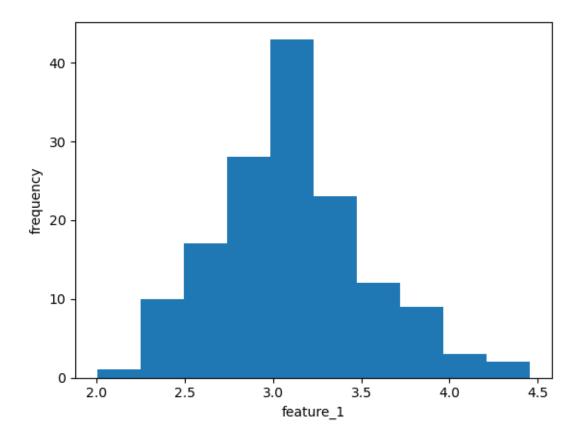
[70]: iris = np.genfromtxt("data/iris.txt", delimiter=None)
  Y = iris[:, -1]
  X = iris[:, 0:-1]
  print(X.shape)

(148, 4)
  # of observations is 148, # of features is 4.

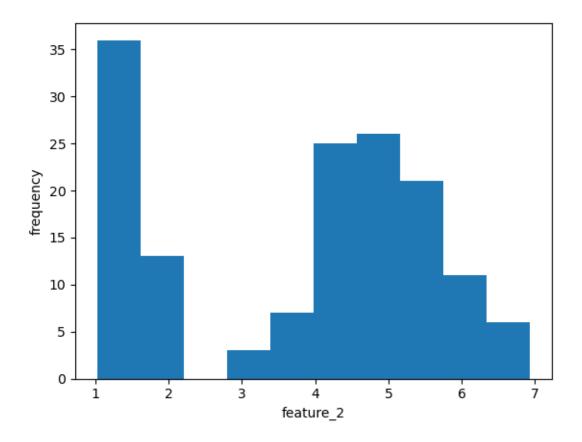
[71]: plt.hist(X[:, 0])
  plt.xlabel('feature_0')
  plt.ylabel('frequency')
  plt.show()
```



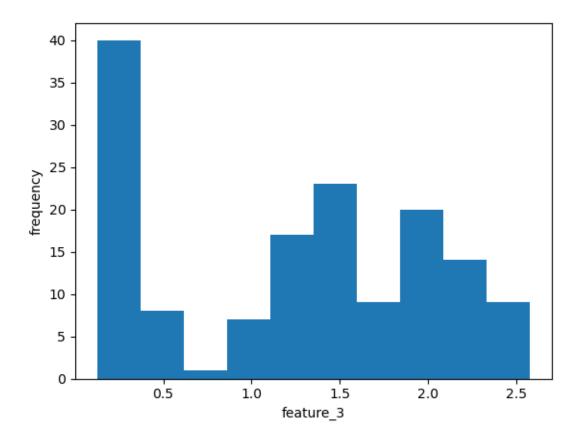
```
[72]: plt.hist(X[:, 1])
   plt.xlabel('feature_1')
   plt.ylabel('frequency')
   plt.show()
```



```
[73]: plt.hist(X[:, 2])
   plt.xlabel('feature_2')
   plt.ylabel('frequency')
   plt.show()
```

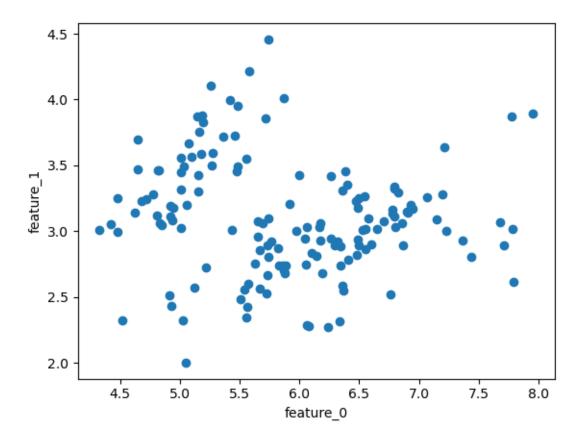


```
[74]: plt.hist(X[:, 3])
   plt.xlabel('feature_3')
   plt.ylabel('frequency')
   plt.show()
```

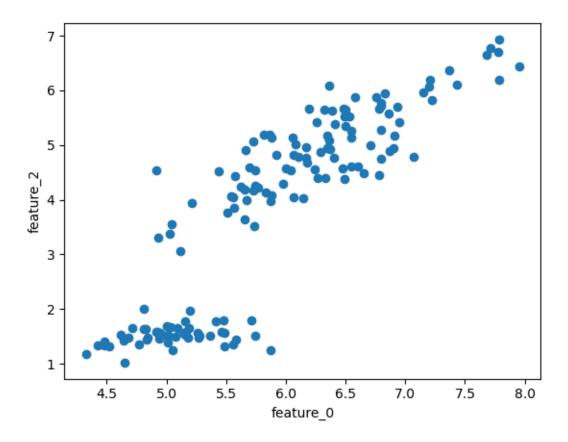


```
Statistics for feature 0: 5.9 0.833
Statistics for feature 1: 3.099 0.436
Statistics for feature 2: 3.82 1.754
Statistics for feature 3: 1.253 0.759
```

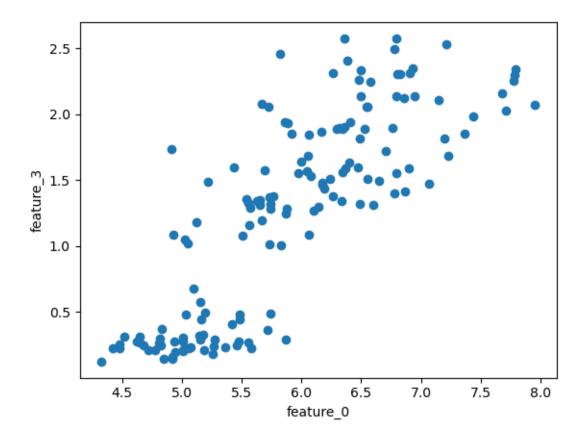
```
[76]: plt.scatter(X[:, 0], X[:, 1])
   plt.xlabel('feature_0')
   plt.ylabel('feature_1')
   plt.show()
```



```
[77]: plt.scatter(X[:, 0], X[:, 2])
   plt.xlabel('feature_0')
   plt.ylabel('feature_2')
   plt.show()
```



```
[78]: plt.scatter(X[:, 0], X[:, 3])
   plt.xlabel('feature_0')
   plt.ylabel('feature_3')
   plt.show()
```

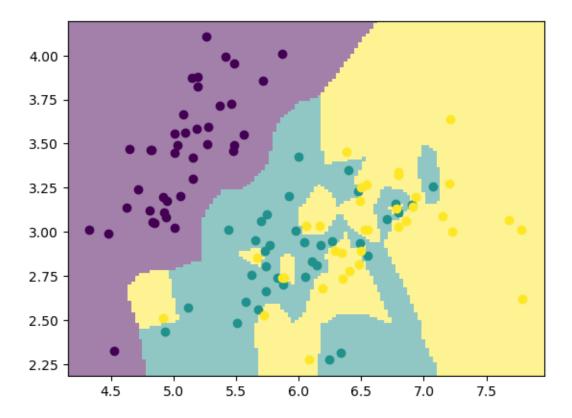


Problem 2: kNN predictions

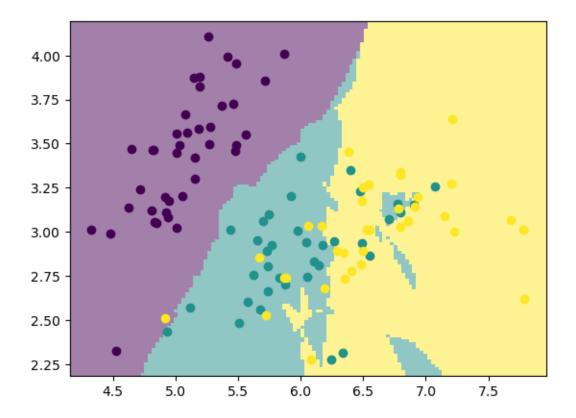
```
[86]: import mltools as ml

X, Y = ml.shuffleData(X, Y)
Xtr, Xva, Ytr, Yva = ml.splitData(X[:, 0:2], Y, 0.75)
knn = ml.knn.knnClassify()

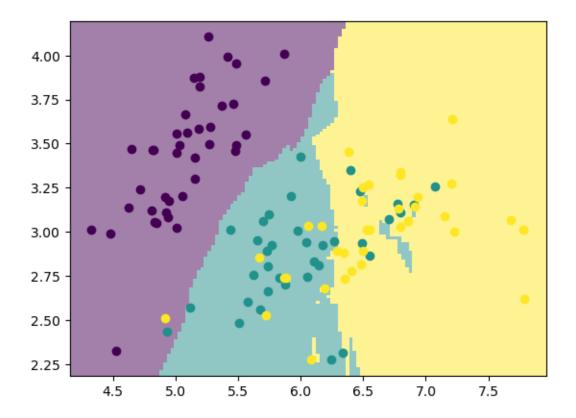
[80]: knn.train(Xtr, Ytr, 1)
YvaHat = knn.predict(Xva)
ml.plotClassify2D(knn, Xtr, Ytr)
```



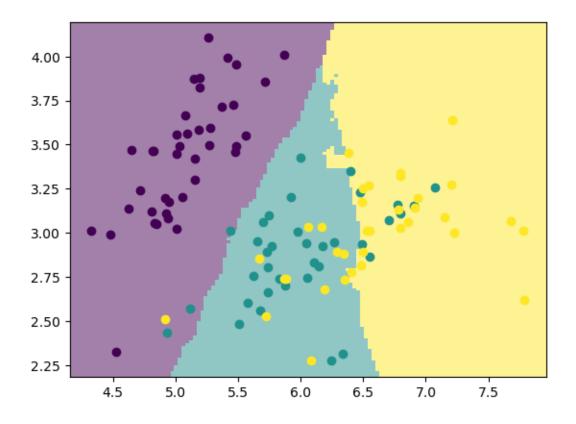
[81]: knn.train(Xtr, Ytr, 5)
YvaHat = knn.predict(Xva)
ml.plotClassify2D(knn, Xtr, Ytr)



```
[82]: knn.train(Xtr, Ytr, 10)
    YvaHat = knn.predict(Xva)
    ml.plotClassify2D(knn, Xtr, Ytr)
```



[83]: knn.train(Xtr, Ytr, 50)
YvaHat = knn.predict(Xva)
ml.plotClassify2D(knn, Xtr, Ytr)

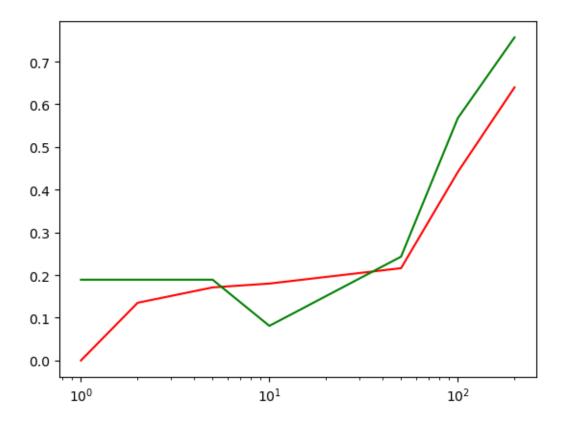


```
[84]: K = [1, 2, 5, 10, 50, 100, 200];
errTrain = np.zeros(shape=len(K))

for i, k in enumerate(K):
    learner = ml.knn.knnClassify(Xtr, Ytr, k)
    Yhat = learner.predict(Xtr)
    errTrain[i] = np.count_nonzero(Yhat != Ytr) / len(Yhat)
    Yhat = learner.predict(Xva)
    errValid[i] = np.count_nonzero(Yhat != Yva) / len(Yhat)

plt.semilogx(K, errTrain, color='r')
plt.semilogx(K, errValid, color='g')
```

[84]: [<matplotlib.lines.Line2D at 0x11ed900d0>]



I recommended k=10, because it produces the lowest validation error.

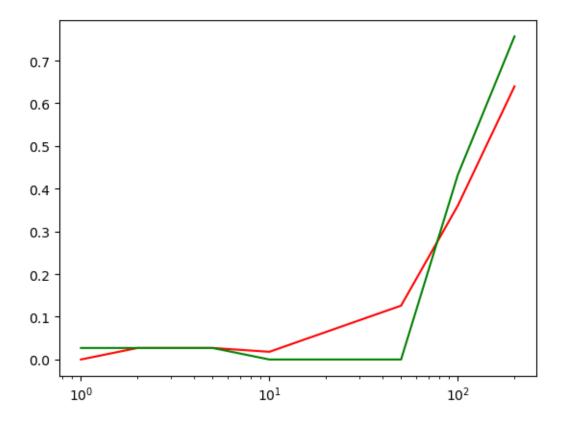
```
[85]: Xtr, Xva, Ytr, Yva = ml.splitData(X, Y, 0.75)

K = [1, 2, 5, 10, 50, 100, 200];
errTrain = np.zeros(shape=len(K))
errValid = np.zeros(shape=len(K))

for i, k in enumerate(K):
    learner = ml.knn.knnClassify(Xtr, Ytr, k)
    Yhat = learner.predict(Xtr)
    errTrain[i] = np.count_nonzero(Yhat != Ytr) / len(Yhat)
    Yhat = learner.predict(Xva)
    errValid[i] = np.count_nonzero(Yhat != Yva) / len(Yhat)

plt.semilogx(K, errTrain, color='r')
plt.semilogx(K, errValid, color='g')
```

[85]: [<matplotlib.lines.Line2D at 0x12d6f7e50>]



The plots are similar, when k is very large, both training and validation errors increased. The recommendation is still k=10 since it has both reasonable training and validation errors. For k=50, the training error increases, which makes it start to have underfitting problem.

0.0.1 Statement of Collaboration

This homework was done completely by Jiachen Sun without collabration.

[]: