

question1

February 28, 2020

```
[1]: # libraries
import numpy as np
import pandas as pd
import random
import seaborn as sns; sns.set()
from sklearn import neighbors
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
import sklearn.preprocessing
import matplotlib.pyplot as plt

[2]: #question 1
#Columns/Features
D = ['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
     ↪ 'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density', 'pH',
     ↪ 'sulphates', 'alcohol']
L = 'quality'
C = 'color'
DL = D + [L]
DC = D + [C]
DLC = DL + [C]

#Loading Data set
wine_r = pd.read_csv("winequality-red.csv", sep=';')
#Loading Data set
wine_w = pd.read_csv("winequality-white.csv", sep=';')
wine_w = wine_w.copy()
wine_w[C] = np.zeros(wine_w.shape[0])
wine_r[C] = np.ones(wine_r.shape[0])
wine = pd.concat([wine_w, wine_r])

[3]: g = sns.pairplot(wine[DC], vars=D, hue="color") #pairplot without normalization
```



```
[4]: from sklearn.preprocessing import StandardScaler
zScore=StandardScaler()
zScore.fit(wine[D])
wine[D]=zScore.transform(wine[D])
g = sns.pairplot(wine[DC],vars= D ,hue="color")
```

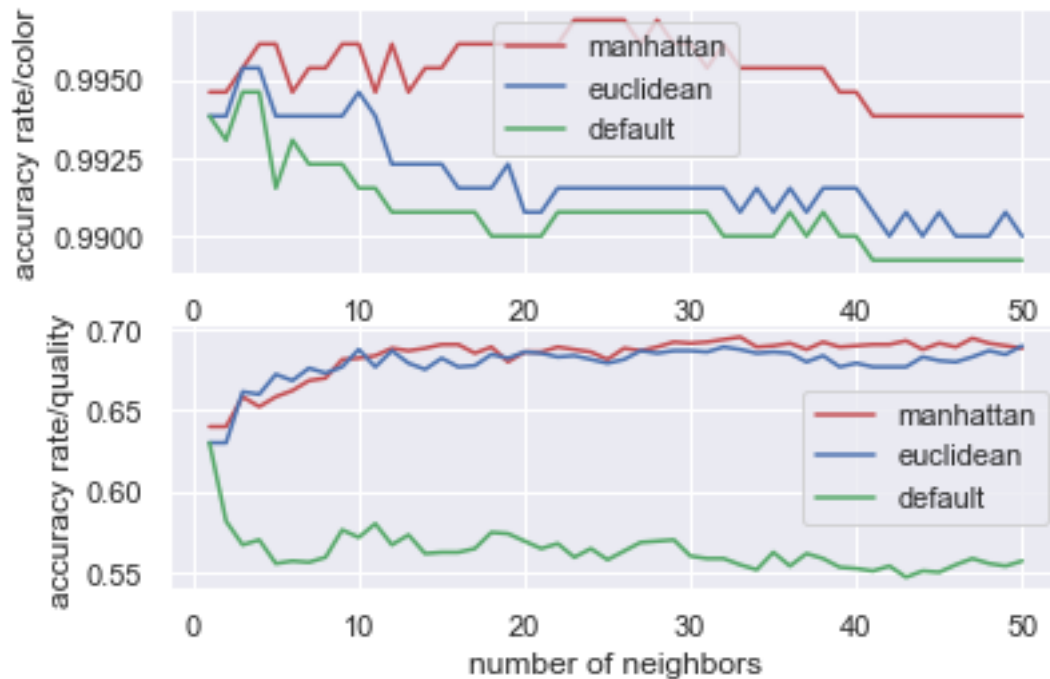


```
[5]: from sklearn.neighbors import KNeighborsClassifier
X = wine[D].values
y1 = np.ravel(wine[[C]])
y2 = np.ravel(wine[[L]])
ran = 42
X_train, X_test, y1_train, y1_test = train_test_split(X, y1, test_size=0.2,
↳random_state = ran)
X_train, X_test, y2_train, y2_test = train_test_split(X, y2, test_size=0.2,
↳random_state = ran)
neighbors=range(1,51)
result1=[]
result2=[]
result3=[]
plt.subplot(2,1,1)
```

```

for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y1_train)
    neigh_2.fit(X_train,y1_train)
    neigh_3.fit(X_train,y1_train)
    result1.append(neigh_1.score(X_test,y1_test))
    result2.append(neigh_2.score(X_test,y1_test))
    result3.append(neigh_3.score(X_test,y1_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/color')
plt.plot(neighbors, result1,'r',label="manhattan")
plt.plot(neighbors, result2,'b',label="euclidean")
plt.plot(neighbors, result3,'g',label="default")
plt.legend()
plt.subplot(2,1,2)
result1=[]
result2=[]
result3=[]
for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y2_train)
    neigh_2.fit(X_train,y2_train)
    neigh_3.fit(X_train,y2_train)
    result1.append(neigh_1.score(X_test,y2_test))
    result2.append(neigh_2.score(X_test,y2_test))
    result3.append(neigh_3.score(X_test,y2_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/quality')
plt.plot(neighbors, result1,'r',label="manhattan")
plt.plot(neighbors, result2,'b',label="euclidean")
plt.plot(neighbors, result3,'g',label="default")
plt.legend()
plt.show()

```

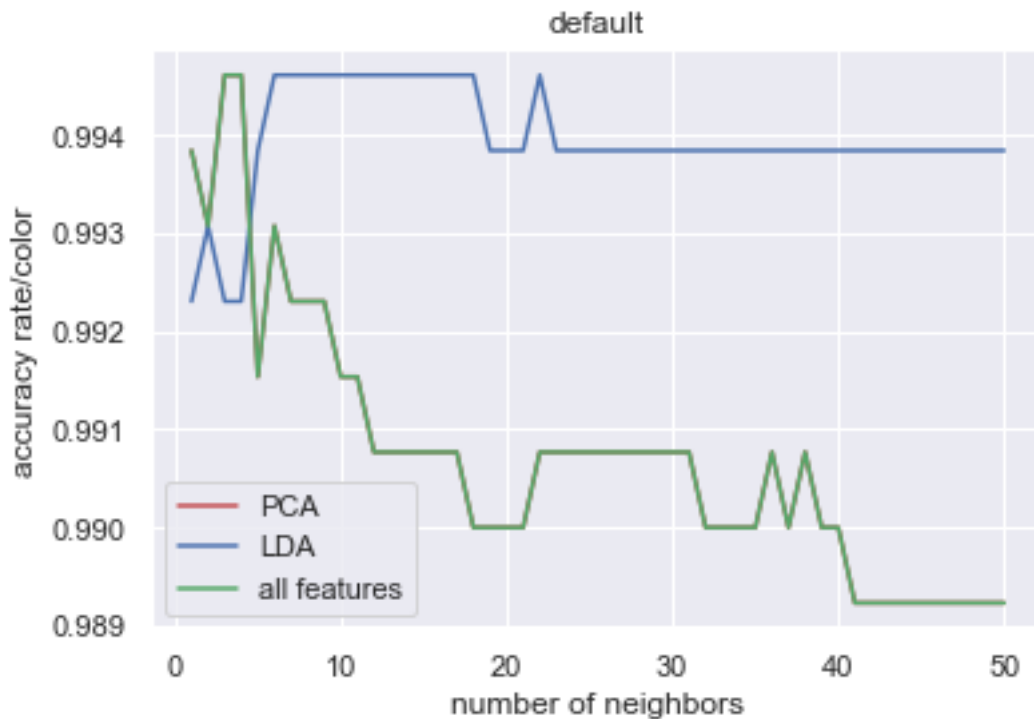


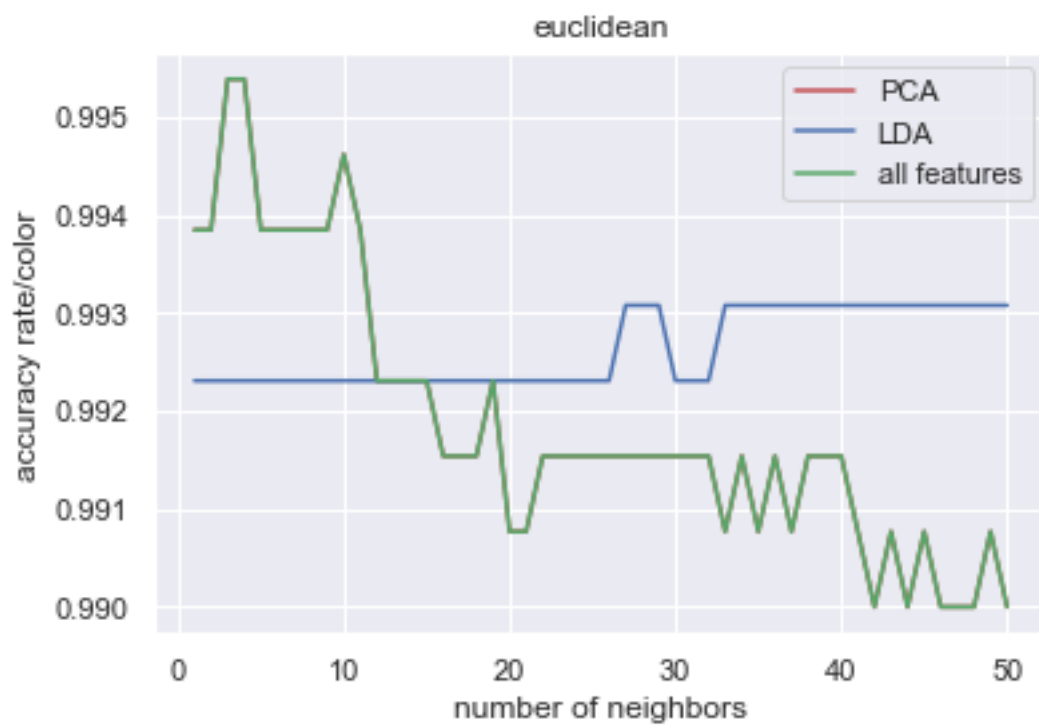
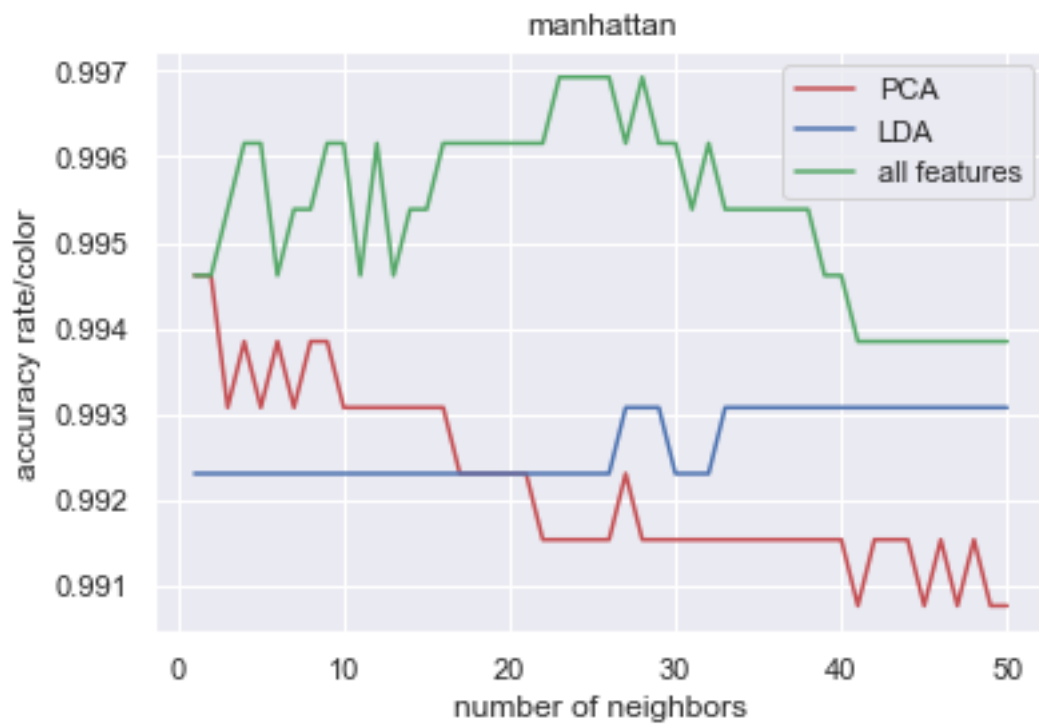
```
[6]: from sklearn.decomposition import PCA
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
from sklearn.neighbors import KNeighborsClassifier
pca=PCA().fit(X_train)
X_train_PCA=pca.transform(X_train)
X_test_PCA=pca.transform(X_test)
lda=LinearDiscriminantAnalysis().fit(X_train,y1_train)
X_train_LDA=lda.transform(X_train)
X_test_LDA=lda.transform(X_test)
for i in range(3):
    result1=[]
    result2=[]
    result3=[]
    for n in neighbors:
        if(i==0):
            neigh_1=KNeighborsClassifier(n_neighbors=n)
            neigh_2 = KNeighborsClassifier(n_neighbors=n)
            neigh_3 = KNeighborsClassifier(n_neighbors=n)
        elif(i==1):
            neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
            neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
            neigh_3=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
        else:
            neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
```

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    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
neigh_1.fit(X_train_PCA,y1_train)
neigh_2.fit(X_train_LDA,y1_train)
neigh_3.fit(X_train,y1_train)
result1.append(neigh_1.score(X_test_PCA,y1_test))
result2.append(neigh_2.score(X_test_LDA,y1_test))
result3.append(neigh_3.score(X_test,y1_test))
if(i==0):
    plt.title("default")
elif(i==1):
    plt.title("manhattan")
else:
    plt.title("euclidean")
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/color')
plt.plot(neighbors, result1,'r',label="PCA")
plt.plot(neighbors, result2,'b',label="LDA")
plt.plot(neighbors, result3,'g',label="all features")
plt.legend()
plt.show()

```





```

[7]: set_1=D
      #set_2 generated by PCA model
      set_3=['residual sugar', 'chlorides', 'total sulfur dioxide', 'density']
      #set_4 will be automaticlly generated by LDA model

[8]: from sklearn.neighbors import KNeighborsClassifier
      X = wine[set_1].values
      y1 = np.ravel(wine[[C]])
      y2 = np.ravel(wine[[L]])
      ran = 42
      X_train, X_test, y1_train, y1_test = train_test_split(X, y1, test_size=0.2,
      ↪random_state = ran)
      X_train, X_test, y2_train, y2_test = train_test_split(X, y2, test_size=0.2,
      ↪random_state = ran)
      neighbors=range(1,50)
      result1=[]
      result2=[]
      result3=[]
      plt.subplot(2,1,1)
      for n in neighbors:
          neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
          neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
          neigh_3=KNeighborsClassifier(n_neighbors=n)
          neigh_1.fit(X_train,y1_train)
          neigh_2.fit(X_train,y1_train)
          neigh_3.fit(X_train,y1_train)
          result1.append(neigh_1.score(X_test,y1_test))
          result2.append(neigh_2.score(X_test,y1_test))
          result3.append(neigh_3.score(X_test,y1_test))
      plt.xlabel('number of neighbors')
      plt.ylabel('accuracy rate/color')
      plt.plot(neighbors, result1,'r',label="manhattan")
      plt.plot(neighbors, result2,'b',label="euclidean")
      plt.plot(neighbors, result3,'g',label="default")
      plt.legend()
      plt.subplot(2,1,2)
      result1=[]
      result2=[]
      result3=[]
      for n in neighbors:
          neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
          neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
          neigh_3=KNeighborsClassifier(n_neighbors=n)
          neigh_1.fit(X_train,y2_train)
          neigh_2.fit(X_train,y2_train)
          neigh_3.fit(X_train,y2_train)
          result1.append(neigh_1.score(X_test,y2_test))

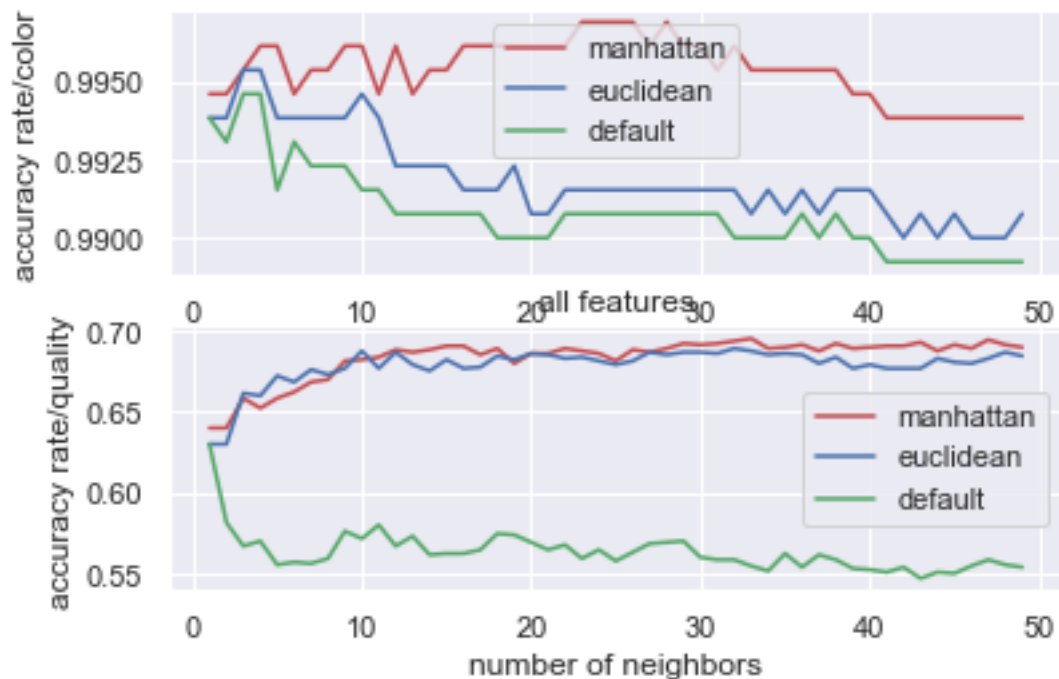
```



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    result2.append(neigh_2.score(X_test,y2_test))
    result3.append(neigh_3.score(X_test,y2_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/quality')
plt.title('all features')
plt.plot(neighbors, result1,'r',label="manhattan")
plt.plot(neighbors, result2,'b',label="euclidean")
plt.plot(neighbors, result3,'g',label="default")
plt.legend()
plt.show()

```



```

[9]: from sklearn.neighbors import KNeighborsClassifier
X = wine[D].values
y1 = np.ravel(wine[[C]])
y2 = np.ravel(wine[[L]])
ran = 42
X_train, X_test, y1_train, y1_test = train_test_split(X, y1, test_size=0.2,
    ↪ random_state = ran)
X_train, X_test, y2_train, y2_test = train_test_split(X, y2, test_size=0.2,
    ↪ random_state = ran)
pca=PCA(n_components=5).fit(X_train)
X_train=pca.transform(X_train)
X_test=pca.transform(X_test)
neighbors=range(1,50)

```

```

result1=[]
result2=[]
result3=[]
plt.subplot(2,1,1)
for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y1_train)
    neigh_2.fit(X_train,y1_train)
    neigh_3.fit(X_train,y1_train)
    result1.append(neigh_1.score(X_test,y1_test))
    result2.append(neigh_2.score(X_test,y1_test))
    result3.append(neigh_3.score(X_test,y1_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/color')
plt.plot(neighbors, result1,'r',label="manhattan")
plt.plot(neighbors, result2,'b',label="euclidean")
plt.plot(neighbors, result3,'g',label="default")
plt.legend()
plt.subplot(2,1,2)
result1=[]
result2=[]
result3=[]
for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y2_train)
    neigh_2.fit(X_train,y2_train)
    neigh_3.fit(X_train,y2_train)
    result1.append(neigh_1.score(X_test,y2_test))
    result2.append(neigh_2.score(X_test,y2_test))
    result3.append(neigh_3.score(X_test,y2_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/quality')
plt.title('first 5 features')
plt.plot(neighbors, result1,'r',label="manhattan")
plt.plot(neighbors, result2,'b',label="euclidean")
plt.plot(neighbors, result3,'g',label="default")
plt.legend()
plt.show()

```

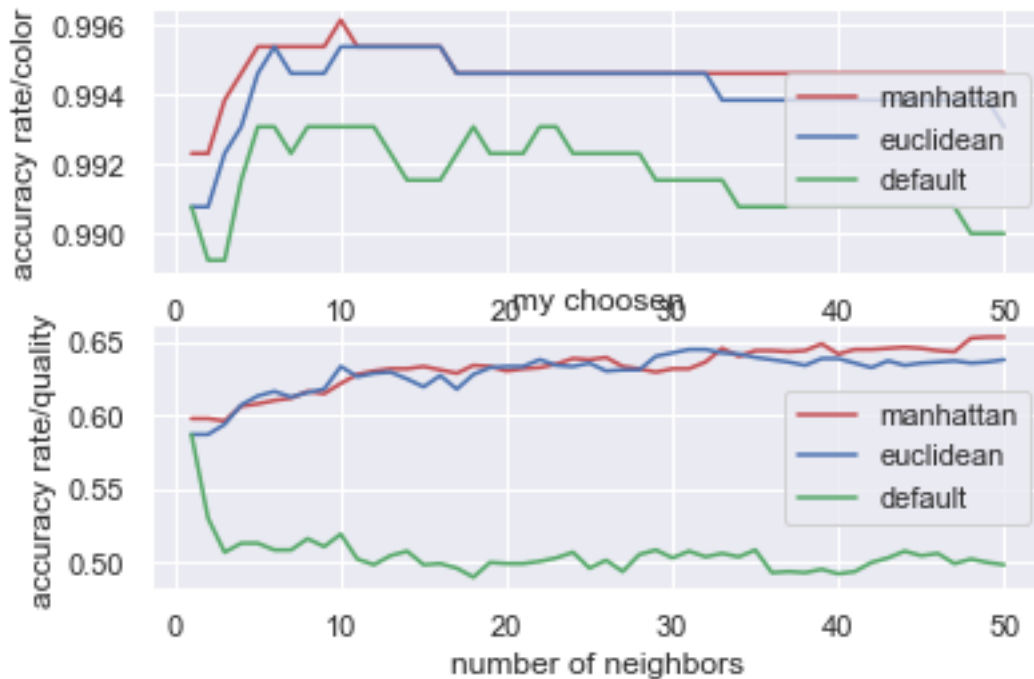


```
[10]: from sklearn.neighbors import KNeighborsClassifier
X = wine[set_3].values
y1 = np.ravel(wine[[C]])
y2 = np.ravel(wine[[L]])
ran = 42
X_train, X_test, y1_train, y1_test = train_test_split(X, y1, test_size=0.2,
↳random_state = ran)
X_train, X_test, y2_train, y2_test = train_test_split(X, y2, test_size=0.2,
↳random_state = ran)
neighbors=range(1,51)
result1=[]
result2=[]
result3=[]
plt.subplot(2,1,1)
for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y1_train)
    neigh_2.fit(X_train,y1_train)
    neigh_3.fit(X_train,y1_train)
    result1.append(neigh_1.score(X_test,y1_test))
    result2.append(neigh_2.score(X_test,y1_test))
    result3.append(neigh_3.score(X_test,y1_test))
```

```

plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/color')
plt.plot(neighbors, result1, 'r', label="manhattan")
plt.plot(neighbors, result2, 'b', label="euclidean")
plt.plot(neighbors, result3, 'g', label="default")
plt.legend()
plt.subplot(2,1,2)
result1=[]
result2=[]
result3=[]
for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y2_train)
    neigh_2.fit(X_train,y2_train)
    neigh_3.fit(X_train,y2_train)
    result1.append(neigh_1.score(X_test,y2_test))
    result2.append(neigh_2.score(X_test,y2_test))
    result3.append(neigh_3.score(X_test,y2_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/quality')
plt.title('my choosen')
plt.plot(neighbors, result1, 'r', label="manhattan")
plt.plot(neighbors, result2, 'b', label="euclidean")
plt.plot(neighbors, result3, 'g', label="default")
plt.legend()
plt.show()

```

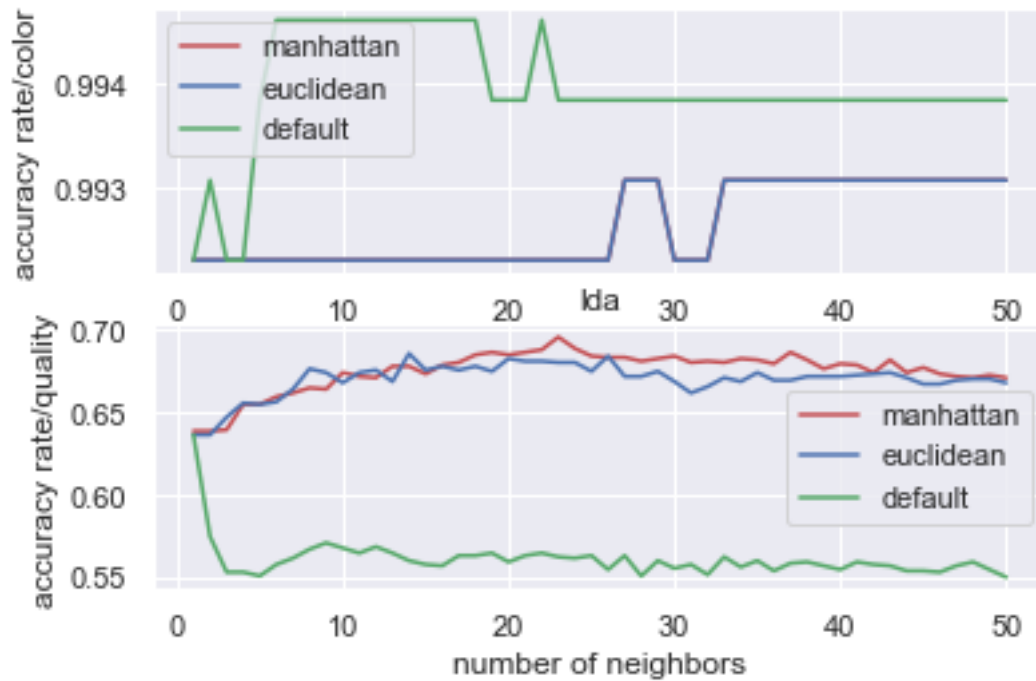


```
[11]: from sklearn.neighbors import KNeighborsClassifier
X = wine[D].values
y1 = np.ravel(wine[[C]])
y2 = np.ravel(wine[[L]])
ran = 42
X_train, X_test, y1_train, y1_test = train_test_split(X, y1, test_size=0.2,
    ↪random_state = ran)
X_train, X_test, y2_train, y2_test = train_test_split(X, y2, test_size=0.2,
    ↪random_state = ran)
lda=LinearDiscriminantAnalysis().fit(X_train,y1_train)
X_train=lda.transform(X_train)
X_test=lda.transform(X_test)
neighbors=range(1,51)
result1=[]
result2=[]
result3=[]
plt.subplot(2,1,1)
for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y1_train)
    neigh_2.fit(X_train,y1_train)
    neigh_3.fit(X_train,y1_train)
```

```

        result1.append(neigh_1.score(X_test,y1_test))
        result2.append(neigh_2.score(X_test,y1_test))
        result3.append(neigh_3.score(X_test,y1_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/color')
plt.plot(neighbors, result1,'r',label="manhattan")
plt.plot(neighbors, result2,'b',label="euclidean")
plt.plot(neighbors, result3,'g',label="default")
plt.legend()
plt.subplot(2,1,2)
X = wine[D].values
y1 = np.ravel(wine[[C]])
y2 = np.ravel(wine[[L]])
X_train, X_test, y1_train, y1_test = train_test_split(X, y1, test_size=0.2,
↳random_state = ran)
X_train, X_test, y2_train, y2_test = train_test_split(X, y2, test_size=0.2,
↳random_state = ran)
lda=LinearDiscriminantAnalysis().fit(X_train,y2_train)
X_train=lda.transform(X_train)
X_test=lda.transform(X_test)
result1=[]
result2=[]
result3=[]
for n in neighbors:
    neigh_1=KNeighborsClassifier(n_neighbors=n,weights='distance',p=1)
    neigh_2=KNeighborsClassifier(n_neighbors=n,weights='distance',p=2)
    neigh_3=KNeighborsClassifier(n_neighbors=n)
    neigh_1.fit(X_train,y2_train)
    neigh_2.fit(X_train,y2_train)
    neigh_3.fit(X_train,y2_train)
    result1.append(neigh_1.score(X_test,y2_test))
    result2.append(neigh_2.score(X_test,y2_test))
    result3.append(neigh_3.score(X_test,y2_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/quality')
plt.title('lda')
plt.plot(neighbors, result1,'r',label="manhattan")
plt.plot(neighbors, result2,'b',label="euclidean")
plt.plot(neighbors, result3,'g',label="default")
plt.legend()
plt.show()

```



```
[12]: pca=PCA(n_components=2).fit(wine[D])
      X1=pd.DataFrame(pca.transform(wine[D]))
      X1.columns=['feature1','feature2']
      X1['color']=wine[C].values
      X1['quality']=wine[L].values
      g = sns.scatterplot(x="feature1", y="feature2",
                          hue="color", data=X1)
```



```
[13]: pca=PCA(n_components=2).fit(wine[D])
X1=pd.DataFrame(pca.transform(wine[D]))
X1.columns=['feature1','feature2']
X1['color']=wine[C].values
X1['quality']=wine[L].values
g = sns.scatterplot(x="feature1", y="feature2",
                    hue="quality", data=X1)
```




```
[14]: lda=LinearDiscriminantAnalysis(n_components=2).fit(wine[D],wine[C])
X2=lda.transform(wine[D])
X2=np.column_stack((X2,[0 for i in range(len(X2))]))
X2=pd.DataFrame(X2)
X2.columns=['feature1','feature2']
X2['color']=wine[C].values
g = sns.scatterplot(x="feature1", y="feature2",
                    hue="color", data=X2,alpha=0.5)

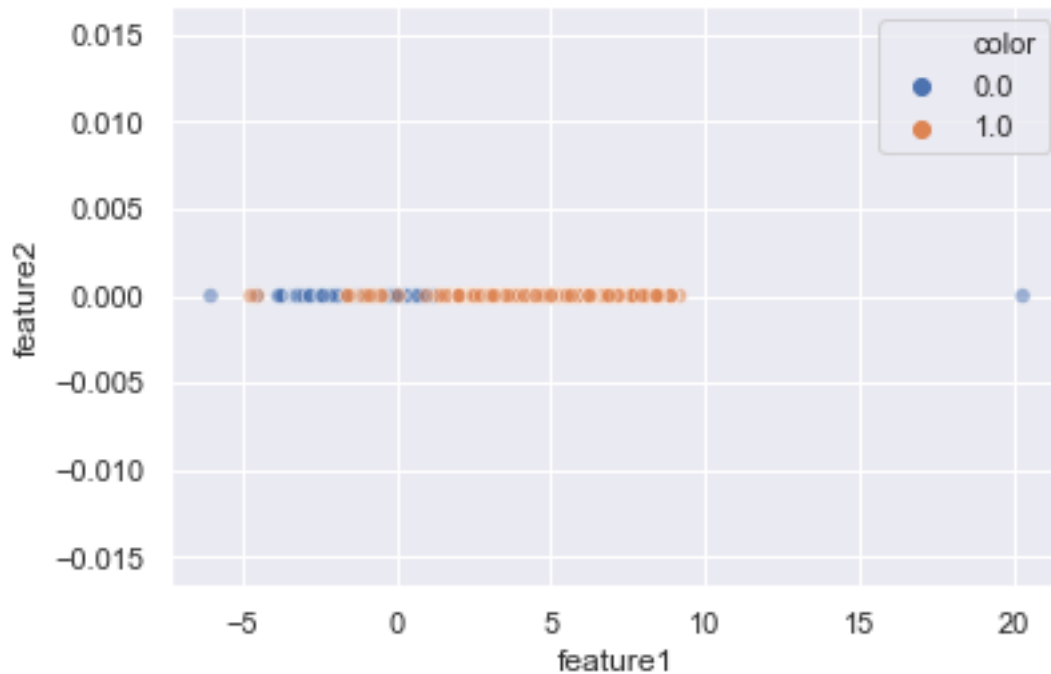
#g = sns.pairplot(X2, height=3,vars=['feature1','feature2'],hue='quality')
```

/usr/local/lib/python3.7/site-packages/sklearn/discriminant_analysis.py:463:
 ChangedBehaviorWarning: n_components cannot be larger than min(n_features,
 n_classes - 1). Using min(n_features, n_classes - 1) = min(11, 2 - 1) = 1
 components.

ChangedBehaviorWarning)

/usr/local/lib/python3.7/site-packages/sklearn/discriminant_analysis.py:469:
 FutureWarning: In version 0.23, setting n_components > min(n_features, n_classes
 - 1) will raise a ValueError. You should set n_components to None (default), or
 a value smaller or equal to min(n_features, n_classes - 1).

warnings.warn(future_msg, FutureWarning)



```
[15]: lda=LinearDiscriminantAnalysis(n_components=2).fit(wine[D],wine[L])
X2=lda.transform(wine[D])
X2=pd.DataFrame(X2)
X2.columns=['feature1','feature2']
X2['quality']=wine[L].values
g = sns.scatterplot(x="feature1", y="feature2",size='quality',
                    hue="quality", data=X2)
```



bonus 2 code

```
[16]: f=['volatile acidity', 'total sulfur dioxide', 'density', 'alcohol']
result1=[]
result2=[]
X1_train, X1_test, y1_train, y1_test = train_test_split(wine[D], y2,
↳test_size=0.2, random_state = ran)
X2_train, X2_test, y2_train, y2_test = train_test_split(wine[f], y2,
↳test_size=0.2, random_state = ran)
for n in neighbors:
    result1.append(KNeighborsClassifier(n_neighbors=n).fit(X1_train,y1_train).
↳score(X1_test,y1_test))
    result2.append(KNeighborsClassifier(n_neighbors=n).fit(X2_train,y2_train).
↳score(X2_test,y2_test))
plt.xlabel('number of neighbors')
plt.ylabel('accuracy rate/color')
plt.title('bonus 2')
plt.plot(neighbors, result1,'r',label="all features")
plt.plot(neighbors, result2,'b',label="4 selected features")
plt.legend()
plt.show()
```

