

Experiment 7,8 Date:2/4/2025

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```
#Distance weighted KN regression for predicting selling price in Boston
#House price data. We are not going to use KNeighborsRegressor from sklearn

import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
#read the data
data=pd.read_csv("BostonHousing.csv",header='infer').values

#separt input and target/output part of data
X=data[:,0:-1]
y=data[:,1]

test_split=float(input("Enter a number between 0 to 1 to specify how
much data is required:"))
#split the data
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=test_spli
t)

dist=np.zeros(shape=X_train.shape[0])
pred=np.zeros(shape=X_test.shape[0])

#for storing prediction

#ask the user about number of nearest neighbors to be used ,i.e. k
k=int(input("Enter the number of nearest neighbours to be used,i.e. k:
"))

for i in range(X_test.shape[0]):
    dist=np.sqrt(np.sum((X_train-X_test[i])**2,axis=1)) #calc
    Euclidean dist
    #between currnt test record and all training records
    kminind=np.argpartition(dist,k)[0:k] #finfing indices of k minimum
    distances
    invdist=1/(dist+10e-20)
    denom=sum(invdist[kminind]) #for weight normalisation
    pred[i]=np.dot(invdist[kminind]/denom,y_train[kminind])
    #print(pred.shape)
    #print(kminind)
    #print(pred)
#UDF to calc MAE
def MAE(pred,y_test):
    return np.mean(abs(pred-y_test))

#UDF to calc MSE
def MSE(pred,y_test):
    return np.mean((pred-y_test)**2)
```

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#UDF to calc MAPE
def MAPE(pred,y_test):
    return np.mean(abs((pred-y_test)/y_test))

#calc performance measures mae=MAE(pred,y_test) mse=MSE(pred,y_test)
rmse=np.sqrt(mse) mape=MAPE(pred,y_test) print(mae) print(mse) print(rmse)
print(mape)

#NOW using sklearn
from sklearn.neighbors import KNeighborsRegressor
regressor
from sklearn.metrics import mean_absolute_error, mean_squared_error
#performance m

#creatign instance opf class #class for KNN
model=KNeighborsRegressor(n_neighbors=k, weights='distance')
#training model
model.fit(X_train,y_train)
#using the trained model for making prediction
pred=model.predict(X_test)
#calc performance measure
mae=mean_absolute_error(y_test,pred)
mse=mean_squared_error(y_test,pred)
print("Using Sklear:\nMAE:",mae)
print("MSE:",mse)

```

Enter a number between 0 to 1 to specify how much data is required:

0.2

Enter the number of nearest neighbours to be used,i.e. k:

4.505199443581134

48.64868125211801

6.974860661842501

0.19504709173038556

5

Using Sklear:

MAE: 4.505199443581133

MSE: 48.64868125211801

#EXPERIMENT 8

#clustering iris flower using k-means and without sklearn for k-means

#we will use Euclidean distance as the distance measure

```
import pandas as pd
```

```

import numpy as np
from sklearn.model_selection import train_test_split          #to split data
in train and test
from sklearn.metrics import classification_report             #for
classification report

data=pd.read_csv("Iris.csv",header='infer').values
X=data[:,1:-1]
y=data[:, -1]

test_split=float(input("Enter a number between 0 to 1 to specify how
much data is required:"))
#split the data
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=test_spli
t,stratify=y)
#ask the user about number of nearest neighbors to be used ,i.e. k
k=int(input("Enter the number of nearest neighbours to be used,i.e. k:
"))
#number of iterations
n=int(input("Enter the number of iterations you want to run
algorithm:"))

#array to store centroids and final centroids
centroids=np.zeros(shape=(k,X_train.shape[1]))
#for selceting rand train poijsnts as initial centroids
per=np.random.permutation(X_train.shape[0])
#select random train points as init centroids
for i in range(k):

    centroids[i,:]=X_train[per[i],:]

for it in range(n):
    dist=np.zeros(shape=(k,X_train.shape[0]))          #for storing dist bw
each
# pair of centroids and trainig data points
for i in range(k):          #compute these distance
    dist[i,:]=np.sqrt(np.sum((X_train-centroids[i,:])**2,axis=1))
membership=np.argmin(dist,axis=0)          #decide membership for each
training dataponts
for i in range(k): #update centroids before we start with iteration
    centroids[i,:]=np.mean(X_train[membership==i,:],axis=0)
print("Centroids after " + str(n) + "iterations:")
print(centroids)
#now inference time
dist=np.zeros(shape=(k,X_test.shape[0]))
for i in range(k):
    dist[i]=np.sqrt(np.sum(X_test-centroids[i])**2,axis=1)

```

```
membership=np.argmin(dist,axis=0)
```

```
#display predicted membership.
```

```
#class number and cluster number may not be same, so be careful while evaluating
```

```
print(y_test.astype(int))
```

```
print(membership)
```

Enter a number between 0 to 1 to specify how much data is required: .2

Enter the number of nearest neighbours to be used,i.e. k: 3

Enter the number of iterations you want to run algorithm: 80000

Centroids after 80000iterations:

```
[[4.54444444 3.05555556 1.38888889 0.22222222]
 [5.          3.3         1.44117647 0.2
 [6.07978723 2.99042553 4.37553191 1.47021277]]
```


TypeError
last) Traceback (most recent call

Cell In[29], line 47

```
45 dist=np.zeros(shape=(k,X_test.shape[0]))
```

```
46 for i in range(k):
```

```
---> 47     dist[i]=np.sqrt(np.sum(X_test-centroids[i])**2,axis=1)
```

```
48 membership=np.argmin(dist,axis=0)
```

```
50 #display predicted membership.
```

```
51 #class number and cluster number may not be same, so be  
careful while evaluating
```

TypeError: sqrt() got an unexpected keyword argument 'axis'