1.EDA(Exploratory Data Analysis) (We will extract info from our data) In [2]: #Importing libraries import seaborn as sns import pandas as pd import numpy as np import scipy.stats import matplotlib.pyplot as plt In []: # reading testing data X_test=pd.read_csv("https://raw.githubusercontent.com/AammarTufail/machinelearning_ka_chill a/master/Sastaticket_datasets/sastaticket_test.csv") In []: #head X_test.head() In []: # save it to csv file sample (250) X_test.to_csv("sastaticket_test.csv") In []: # reading testing data sample(250) df_test=pd.read_csv("sastaticket_test.csv") df_test.head() In []: # reading training dataset(sample) from github training_data=pd.read_csv("https://raw.githubusercontent.com/AammarTufail/machinelearning_ka _chilla/d4693bff19a9e9d47908b55218daa0b9a77a13c3/Sastaticket_datasets/sastaticket_train.csv" In []: | #head training_data.head(3) In []: # save to csv file training_data.to_csv("sastaticket_trainsample.csv") In [3]: #read df=pd.read_csv("sastaticket_trainsample.csv") In []: #head df.head(2)In []: #shape df.shape In []: # info df.info() In []: #check null values #df.isnull().sum() df.notnull().sum() In []: #summary statistics df.describe() In []: #head df.head(3)In [4]: # first we drop unnamed data df.drop(["Unnamed: 0.3", "Unnamed: 0.2", "Unnamed: 0", "Unnamed: 0.1"], axis=1, inplace=True) In []: #columns df.columns In []: df.nunique() In []: |# separate categorical variables in a list cat_list=["f2", "f3", "f6", "f7", "f8", "f9", "f10"] In []: # for loop to check unique values in each for i in cat_list: print(i, df[i].unique()) print("....")# separatorline In [5]: # As we can see that f2 and f3 have no unique values and they will not effect the target. # f10 is flight no which also has no impact on target. # so it's better to drop f2,f3,f10 df.drop(["f2", "f3", "f10"], axis=1, inplace=True) Assignment#1 If we replace axis from 1 to 0 it will give error of "not found in axis" In [6]: #head df.head() Out[6]: f8 f9 target f1 f6 f7 **0** 2021-01-08 12:43:27.828728+00:00 2021-01-23 05:00:00+00:00 2021-01-23 07:00:00+00:00 gamma True 0.0 0 7400.0 **1** 2021-07-01 04:45:11.397541+00:00 2021-07-01 13:00:00+00:00 2021-07-01 15:00:00+00:00 alpha True 35.0 1 15377.0 **2** 2021-06-24 11:28:47.565115+00:00 2021-07-29 14:00:00+00:00 2021-07-29 16:00:00+00:00 gamma True 20.0 1 6900.0 **3** 2021-06-05 11:09:48.655927+00:00 2021-06-09 16:00:00+00:00 2021-06-09 18:00:00+00:00 alpha True 15.0 1 9707.0 **4** 2021-07-29 09:53:51.065306+00:00 2021-08-23 05:00:00+00:00 2021-08-23 06:55:00+00:00 beta True 20.0 0 6500.0 In []: # casting df.info() In [7]: # as columns f1, f4, f5 shows date and time with dtype object # we convert dtype to datetime #for that import datetime library from datetime import date, time df["f1"]=pd.to_datetime(df["f1"]) df["f4"]=pd.to_datetime(df["f4"]) df["f5"]=pd.to_datetime(df["f5"]) In [8]: #importing date, timedelta from datetime import date, timedelta In [9]: # add a new column(time_to_dparture) by subtracting f1 from f4 $df.insert(0,"time_to_dep(s)" , ((df['f4']-df['f1']).astype('timedelta64[s]')), \ \textit{True})$ # add a new column(travel_time) by subtracting f5 from f4 df.insert(1, "travel_time(s)", ((df["f5"]-df["f4"]).astype("timedelta64[s]")), True) In []: #check head df.head(3)In []: #uniqueness df.nunique() In []: #null values df.notnull().sum() In []: #separating variables cat_col=["f6","f7","f8","f9"] num_col=["time_to_dep(s)", "travel_time(s)"] In []: |# Plotting # take insights through categorical_columns by countplot plt.figure(figsize=(20,45)) **for** i **in** cat_col: plt.subplot(6,3,c) sns.countplot(df[i]) In []: # Plotting # take insights through numerical_columns by disttplot plt.figure(figsize=(20,45)) for 1 in num_col: plt.subplot(6,3,c) sns.distplot(df[i]) c=c+1 In []: # Plotting of target # 1.distplot sns.displot(df["target"]) **Assignment#2** 1.Remove outliers from dataset In [10]: # Detection # IQR Q1 = np.percentile(df['target'], 25, interpolation = 'midpoint') Q3 = np.percentile(df['target'], 75, interpolation = 'midpoint') IQR = Q3 - Q1print("Old Shape: ", df.shape) # Upper bound upper = np.where(df['target'] >= (Q3+1.5*IQR)) # Lower bound lower = np.where(df['target'] <= (Q1-1.5*IQR))</pre> ''' Removing the Outliers ''' df.drop(upper[0], inplace = True) df.drop(lower[0], inplace = True) print("New Shape: ", df.shape) C:\Users\mohsin.DESKTOP-7I5HD4K\AppData\Local\Temp\ipykernel_2396\2468351618.py:3: Deprecatio nWarning: the `interpolation=` argument to percentile was renamed to `method=`, which has add itional options. Users of the modes 'nearest', 'lower', 'higher', or 'midpoint' are encouraged to review the m ethod they. (Deprecated NumPy 1.22) Q1 = np.percentile(df['target'], 25, C:\Users\mohsin.DESKTOP-7I5HD4K\AppData\Local\Temp\ipykernel_2396\2468351618.py:6: Deprecatio nWarning: the `interpolation=` argument to percentile was renamed to `method=`, which has add itional options. Users of the modes 'nearest', 'lower', 'higher', or 'midpoint' are encouraged to review the m ethod they. (Deprecated NumPy 1.22) Q3 = np.percentile(df['target'], 75, Old Shape: (5000, 10) New Shape: (4749, 10) In []: #shape df.shape In []: #Skewness and kurtosis df.skew()

Assignment:

Submitted by:

Submitted to:

Date: (04/08/2022)

Instructions

4. ML model

1. You will have dataset

3. You have to go through EDA

5. Find a sweet spot for cheap ticket

Case Study

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Flight Price Prediction Model

2. Find cheapest and expensive flight at a specific time

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Ahmad is customer of sastaticket.pk.He is planning to fly from karachi to islamabad.

• But he doesnt buy ticket now and he thinks that prices are too much high now and he must wait. • Can you tell ahmad he is making wrong decision with your analysis and a very confident prediction

In []: #Kurtosis df.kurtosis() In []: df.head() In [11]: # encoding df.drop(["f1", "f4", "f5"], axis=1, inplace=True) In [12]: #encoding from sklearn.preprocessing import LabelEncoder le=LabelEncoder() In [13]: # Apply label encoder

In []: #head

Out[16]:

df.head()

df3.head()

0

1

3

In [18]: df3.info()

3

In [20]: | df3.head()

0

1

3

Out[20]:

In [14]: df.to_csv("st_file.csv")

In [16]: df3=pd.read_csv("st_file.csv")

1

3

Column

f6 f7

f8

f9

target

dtypes: float64(3), int64(5)

memory usage: 296.9 KB

0

1

3

In []: ##Shapiro test to check normality

stat, p=shapiro(df["f9"])

print("stat=", stat)

Assignment#3

sc=StandardScaler()

In []: #head

In []: #columns

df.head()

In []: #again normality

print("stat=", stat)

print("p=",p)

if p > 0.05:

df.columns

y=df['target']

ML Modelling

from sklearn.svm import SVR

metrics for regression

metrics for classification

F1,Recall score,precision score

dt=DecisionTreeRegressor()

knn=KNeighborsRegressor()

print(i)

#to save prediction res=pd.DataFrame(pred)

Assignment#4

)), **True**)

)), **True**)

le=LabelEncoder()

In []: #dropping unnecessaary columns

In []: # Apply label encoder

df_test.shape

df_test.head(2)

In []: #final data prediction

df_test.head(2)

In []: #predicted value

pred1

lr=LinearRegression().fit(X,y)

res.index=df_test.index # its important for comparison

pred1=lr.predict([[420706.0,7200.0,3,0,1,1]])

res.to_csv("prediction_results_with_traintestsplit2.csv")

pred=lr.predict(df_test) res=pd.DataFrame(pred)

res.columns=["prediction"]

In []: #encoding

In []: #shape

In []: #head

In []: #head

In []: df.head(2)

res.columns=["prediction"]

we convert dtype to datetime #for that import datetime library from datetime import date, time

In []: # shortening the model names lr=LinearRegression()

svr=SVR()

In []: | # model loop

r2_score,mean_absolute_error,mean_squared_error

for i in [lr,dt,svr,knn]: #read all model

pred=i.predict(X_test) #predict

i.fit(X_train,y_train) #fitting our model

if abs(train_score-test_score) <=0.1:</pre>

test_score=r2_score(y_test, pred) #test score

res.index=X_test.index # its important for comparison

Now apply all modification steps df_test

In []: # as columns f1, f4, f5 shows date and time with dtype object

In []: |# add a new column(time_to_dparture) by subtracting f1 from f4

add a new column(travel_time) by subtracting f5 from f4

from sklearn.preprocessing import LabelEncoder

df_test["f6"]=le.fit_transform(df_test["f6"]) df_test["f7"]=le.fit_transform(df_test["f7"]) df_test["f8"]=le.fit_transform(df_test["f8"])

df_test["f1"]=pd.to_datetime(df_test["f1"]) df_test["f4"]=pd.to_datetime(df_test["f4"]) df_test["f5"]=pd.to_datetime(df_test["f5"])

res.to_csv("prediction_results_with_traintestsplit.csv")

print("R2 score is:",r2_score(y_test,pred))

print("RMSE is:", mean_squared_error(y_test, pred)) print("----")

#rootmeansquarederror

In []: | # split into X, y

print("p=",p)

if p > 0.05:

from scipy.stats import shapiro

print("data is normal")

In []: # #As data is not normal go for scaling

from sklearn.preprocessing import StandardScaler

df[["target"]]=sc.fit_transform(df[["target"]])

If we apply scaling it will give values in minus

stat, p=shapiro(df["travel_time(s)"])

print("data is normal")

print("data is not normal")

df[["time_to_dep(s)"]]=sc.fit_transform(df[["time_to_dep(s)"]]) # df[["travel_time(s)"]]=sc.fit_transform(df[["travel_time(s)"]])

X=df[['time_to_dep(s)','travel_time(s)', 'f6', 'f7', 'f8', 'f9']]

from sklearn.linear_model import LinearRegression from sklearn.tree import DecisionTreeRegressor

from sklearn.neighbors import KNeighborsRegressor from sklearn.model_selection import train_test_split

(rms)=mean_squared_error(y_test,y_pred, squared=False)

In []: # Because our target(y) is numerical variable so we choose regression model/pipeline

from sklearn.metrics import r2_score,mean_absolute_error,mean_squared_error

X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.2, random_state=42)

print("Mean absolute error is:", mean_absolute_error(y_test, pred)) print("Mean squared error is:", mean_squared_error(y_test, pred))

df_test.insert(0,"time_to_dep(s)" , ((df_test['f4']-df_test['f1']).astype('timedelta64[s]'

df_test.insert(1, "travel_time(s)", ((df_test["f5"]-df_test["f4"]).astype("timedelta64[s]"

df_test.drop(["Unnamed: 0","Unnamed: 0.1","f1","f2","f3","f4","f5","f10"], axis=1,inplace=Tr

train_score=r2_score(y_train,i.predict(X_train)) #train score

print("data is not normal")

In [21]: df3.to_csv("st3.file.csv")

Unnamed: 0

df["f6"]=le.fit_transform(df["f6"]) df["f7"]=le.fit_transform(df["f7"]) df["f8"]=le.fit_transform(df["f8"])

Unnamed: 0 time_to_dep(s) travel_time(s) f6 f7 f8 f9

7200.0 2 1 0 0 7400.0

7200.0 0 1 4 1 15377.0

7200.0 2 1 2 1 6900.0

7200.0 0 1 1 1 9707.0

6900.0 1 1 2 0 6500.0

int64

int64

7200 2 1 0 0 7400

7200 0 1 4 1 15377

7200 2 1 2 1 6900

7200 0 1 1 1 9707

6900 1 1 2 0 6500

Non-Null Count Dtype

4749 non-null int64

4749 non-null int64

4749 non-null int64

4749 non-null float64

4749 non-null

4749 non-null

df3["travel_time(s)"]=df3["travel_time(s)"].astype("int64")

time_to_dep(s) 4749 non-null float64 travel_time(s) 4749 non-null float64

In [19]: | df3["time_to_dep(s)"]=df3["time_to_dep(s)"].astype("int64")

Unnamed: 0 time_to_dep(s) travel_time(s) f6 f7 f8 f9 target

df3["target"]=df3["target"].astype("int64")

1268192

29688

3033072

363011

2142368

1268192.0

3033072.0

363011.0

2142368.0

<class 'pandas.core.frame.DataFrame'> RangeIndex: 4749 entries, 0 to 4748 Data columns (total 8 columns):

29688.0