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#1,2) Downloaded and Loaded the given Dataset
import pandas as pd
import numpy as np
#For Plotting
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.preprocessing import LabelEncoder
from sklearn.linear model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import scale
from sklearn.decomposition import PCA
from sklearn.discriminant analysis import LinearDiscriminantAnalysis
from scipy import stats
from IPython.display import display, HTML
import plotly.express as px
df = pd.read csv('Churn Modelling.csv')
df.head()
#3) a) UNIVARIATE ANALYSIS:
plt.hist(df['Age'])
sns.countplot(df['Age'])
sns.set(rc={'figure.figsize':(22,5)})
sns.countplot(df['Geography'])
#3) b) BIVARIATE ANALYSIS:
plt.scatter(df['EstimatedSalary']
df['Age'])sns.heatmap(df.corr(), annot = True)
plt.scatter( df['Balance'], df['NumOfProducts'])
plt.xlabel('Balance')
plt.ylabel('NumofProducts')
sns.countplot(data=df,x='Geography', hue='Gender')
sns.set(rc={'figure.figsize':(10,5)})
sns.countplot(data=df, x='HasCrCard', hue='Gender')
#3)c) MULTIVARIATE ANALYSIS
sns.pairplot(data = df)
#4) DESCRIPTIVE ANALYSIS
df.describe()
df.info()
df.count()
#5) Handling Missing Values
df.isnull().sum()
#6) Finding Outliers and replacing the outliers
sns.boxplot(df['Age'])
sns.boxplot(df['CreditScore'])
df.columns
Q1=df['Age'].quantile(0.25)
Q3=df['Age'].quantile(0.75)
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IQR=Q3-Q1
whisker width = 1.5
age outliers = df[(df['Age'] < Q1 - whisker width*IQR) | (df['Age'] > Q3)
+ whisker width*IQR)]
age outliers.count()
import numpy as np
lower whisker = Q1 -(whisker width*IQR)
upper whisker = Q3 + (whisker width*IQR)
df['Age']=np.where(df['Age']>upper whisker,upper whisker,np.where(df['Age
'] < lower_whisker, lower_whisker, df['Age']))
sns.boxplot(df['Age'],data=df)
# CreditScore
Q1=df['CreditScore'].quantile(0.25)
Q3=df['CreditScore'].quantile(0.75)
IQR=Q3-Q1
whisker width = 1.5
credscore outliers = df[(df['CreditScore'] < Q1 - whisker width*IQR) |</pre>
(df['CreditScore'] > Q3 + whisker width*IQR)]
credscore outliers.count()
lower whisker = Q1 -(whisker width*IQR)
upper whisker = Q3 + (whisker width*IQR)
df['CreditScore']=np.where(df['CreditScore']>upper whisker,upper whisker,
np.where(df['CreditScore']<lower whisker,lower whisker,df['CreditScore'])</pre>
sns.boxplot(df['CreditScore'])
df['CreditScore'].count()
#7) Checking for categorical columns:
df.info()
from sklearn.preprocessing import LabelEncoder
enc = LabelEncoder()
df['Geography'] = enc.fit transform(df['Geography'])
df['Gender'] = enc.fit transform(df[['Gender']])
df['Geography'].unique(), df['Gender'].unique()
df.head()
#8) SPLITTING DEPENDENT AND INDEPENDENT VARIABLES:
X = df.iloc[:, 3:-1]
X.head()
y = df.iloc[:,-1]
#9) Scaling the Independent Variables:
from sklearn.preprocessing import StandardScaler
scale = StandardScaler()
X = scale.fit transform(X)
#10) Splitting the test and train variables
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X,y ,random state=1,
shuffle=True, train size = 0.8)
X train.shape, X test.shape, y train.shape, y test.shape
from sklearn import linear model
reg = linear model.LogisticRegression()
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reg.fit(X_train, y_train)
y_pred = reg.predict(X_test)
rom sklearn.metrics import classification_report
print(classification_report(y_test, y_pred))