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1 ### md
2 IMPORTING REQUIRED LIBRARIES
3 ###
4 import pandas as pd
5 ###
6 import numpy as np
7 ###
8 import matplotlib.pyplot as plt
9 ###
10 import seaborn as sns
11 ###
12 train = pd.read_csv("train.csv")
13 ###
14 test = pd.read_csv("test.csv")
15 ### md
16 #Viewing data and different features
17
18 ###
19 train.head()
20 ### md
21 train.shape
22 ###
23 train.columns
24 ###
25 train["Sex"].value_counts()
26 ### raw
27 Data Visualization
28 ###
29 #visualizing survivals based on gender
30 train["Died"] = 1 - train["Survived"]
31 train.groupby("Sex").agg("sum")[["Survived", "Died"]].
   plot(kind="bar",figsize=(10,5),stacked=True)
32 ###
33 #Visualizing survivals based on fare
34 figure = plt.figure(figsize=(16,7))
35 plt.hist([train[train["Survived"] == 1]["Fare"],train
   [train["Survived"] == 0]["Fare"],stacked=True, bins
   = 50, label = ['Survived', 'Dead'])
36 plt.xlabel("Fare")
37 plt.ylabel("Number of passengers")
38 plt.legend()
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39 ### md
40 Processing training data
41
42 ###
43 #Cleaning the data by removing irrelevant columns
44 df1 = train.drop(["Name","Ticket","Cabin","
    PassengerId","Died"],axis=1)
45 df1.head(100)
46 ###
47 df1.isnull().sum()
48 ###
49 #converting the categorical features "sex" and "
    Embarked" into numerical values 0 and 1
50 df1.Sex = df1.Sex.map({"female":0,"male":1})
51 df1.Embarked=df1.Embarked.map({"S":0,"C":1,"Q":2,"nan
    ":"NaN"})
52 df1.head()
53 ###
54 #mean age of each sex
55 mean_age_men = df1[df1["Sex"]==1]["Age"].mean()
56 mean_age_women = df1[df1["Sex"]==0]["Age"].mean()
57 ###
58 #Filling all the null values in 'Age' with respective
    mean age
59 df1.loc[(df1.Age.isnull()) & (df1['Sex']==0),'Age']=
    mean_age_women
60 df1.loc[(df1.Age.isnull()) & (df1['Sex']==1),'Age']=
    mean_age_men
61 ###
62 #let's check for the Null values again now
63 df1.isnull().sum()
64 ###
65 #since there exist 2 null values in the embarked
    column, let's drop ththose rows containing null values
66 df1.dropna(inplace=True)
67 ###
68 df1.isnull().sum()
69 ###
70 #Doing Feature Scaling to standardize the independent
    features present in the data in a fixed range
71 df1.Age = (df1.Age-min(df1.Age))/(max(df1.Age)-min(

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71 df1.Age))
72 df1.Fare = (df1.Fare-min(df1.Fare))/(max(df1.Fare)-
    min(df1.Fare))
73 df1.describe()
74 ### md
75 Creating model
76
77 ###
78 #Splitting the data for training and testing
79 from sklearn.model_selection import train_test_split
80 X_train, X_test, y_train, y_test = train_test_split(
81     df1.drop(['Survived'], axis=1),
82     df1.Survived,
83     test_size= 0.2,
84     random_state=0,
85     stratify=df1.Survived)
86 ### md
87 Logistic Regression
88 ###
89 from sklearn.linear_model import LogisticRegression
90 lrmod = LogisticRegression()
91 lrmod.fit(X_train, y_train)
92 from sklearn.metrics import accuracy_score
93 y_predict = lrmod.predict(X_test)
94 accuracy_score(y_test, y_predict)
95 ###
96 #Confusion Matrix
97 from sklearn.metrics import confusion_matrix
98 cma=confusion_matrix(y_test, y_predict)
99 sns.heatmap(cma,annot=True)
100 ### md
101 Processing test data
102 ###
103 #Viewing test data
104 test.head()
105 ###
106 #Cleaning the data by removing irrelevant columns
107 df2=test.drop(['PassengerId', 'Name', 'Ticket', 'Cabin'
    ], axis=1)
108 df2
109 ###

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110 #Converting the categorical features 'Sex' and 'Embarked' into numerical values 0 & 1
111 df2.Sex=df2.Sex.map({'female':0, 'male':1})
112 df2.Embarked=df2.Embarked.map({'S':0, 'C':1, 'Q':2, 'nan': 'NaN'})
113 df2.head()
114 ###
115 #Let's check for the null values
116 df2.isnull().sum()
117 ###
118 #Finding mean age
119 mean_age_men2=df2[df2['Sex']==1]['Age'].mean()
120 mean_age_women2=df2[df2['Sex']==0]['Age'].mean()
121 ###
122 #Filling all the null values in 'Age' and 'Fare' with respective mean age and mean fare
123 df2.loc[(df2.Age.isnull()) & (df2['Sex']==0), 'Age']=mean_age_women2
124 df2.loc[(df2.Age.isnull()) & (df2['Sex']==1), 'Age']=mean_age_men2
125 df2['Fare']=df2['Fare'].fillna(df2['Fare'].mean())
126 ###
127 df2.isnull().sum()
128 ###
129 #Doing Feature Scaling to standardize the independent features present in the data in a fixed range
130 df2.Age = (df2.Age-min(df2.Age))/(max(df2.Age)-min(df2.Age))
131 df2.Fare = (df2.Fare-min(df2.Fare))/(max(df2.Fare)-min(df2.Fare))
132 df2.describe()
133 ### md
134 Prediction
135 ###
136 prediction = lrmod.predict(df2)
137 prediction
138 ###
139 prediction
140 ###
141 submission = pd.DataFrame({"PassengerId": test["

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141 PassengerId"],
142                                     "Survived": prediction})
143 submission.to_csv('submission.csv', index=False)
144 #%%
145 prediction_df = pd.read_csv('submission.csv')
146 #%%
147 sns.countplot(x='Survived', data=prediction_df)
148 #%%
149 sns.countplot(x='Survived', data=prediction_df)sns.
    countplot(x='Survived', data=prediction_df)
150 #%%
151
```