```
\hbox{import numpy as np}\\
import pandas as pd
import seaborn as sns
from matplotlib import pyplot as plt
# !pip install missingno
import missingno as msno
from datetime import date
from sklearn.metrics import accuracy_score
from sklearn.model_selection import train_test_split
from sklearn.neighbors import LocalOutlierFactor
from \ sklearn.preprocessing \ import \ MinMaxScaler, \ Label Encoder, \ Standard Scaler, \ Robust Scaler
from google.colab import files
uploaded = files.upload()
    Choose Files titanic.csv
     • titanic.csv(text/csv) - 61194 bytes, last modified: 6/28/2024 - 100% done
     Saving titanic.csv to titanic.csv
def load():
    data = pd.read_csv("titanic.csv")
    return data
#See the shape of smaller dataset
df = load()
print(df.shape)
→ (891, 12)
sns.boxplot(x=df["Age"])
plt.show()
\overrightarrow{\Rightarrow}
                                                              00 000
                 10
                         20
                                 30
                                         40
                                                         60
                                                                 70
                                                                         80
                                        Age
```

```
q1 = df["Age"].quantile(0.25)
q3 = df["Age"].quantile(0.75)
iqr = q3 - q1
up = q3 + 1.5 * iqr
low = q1 - 1.5 * iqr
```

#Now see outliers
print(df[(df["Age"] < low) | (df["Age"] > up)])

| <u> </u> |     | PassengerId  | Survivad  | Delace | Name \                         |
|----------|-----|--------------|-----------|--------|--------------------------------|
| ~        |     | rassenger tu | Jul VIVEU | LCTG33 | Name v                         |
|          | 33  | 34           | 0         | 2      | Wheadon, Mr. Edward H          |
|          | 54  | 55           | 0         | 1      | Ostby, Mr. Engelhart Cornelius |
|          | 96  | 97           | 0         | 1      | Goldschmidt, Mr. George B      |
|          | 116 | 117          | 0         | 3      | Connors, Mr. Patrick           |

```
280
                  281
                               0
                                                               Duane, Mr. Frank
     456
                  457
                               0
                                       1
                                                     Millet, Mr. Francis Davis
     493
                  494
                               0
                                                       Artagaveytia, Mr. Ramon
                                          Barkworth, Mr. Algernon Henry Wilson
     672
                  673
                                                   Mitchell, Mr. Henry Michael
                                                  Crosby, Capt. Edward Gifford
                               0
     745
                  746
                                       1
     851
                  852
                                                            Svensson, Mr. Johan
           Sex
                 Age
                      SibSp
                             Parch
                                         Ticket
                                                    Fare Cabin Embarked
     33
          male
                66.0
                           0
                                  0
                                     C.A. 24579
                                                 10.5000
                                                            NaN
                                                                       ς
     54
          male
                65.0
                           0
                                  1
                                         113509
                                                 61.9792
                                                            B30
                                                                       C
     96
          male
                71.0
                           0
                                  0
                                       PC 17754
                                                 34.6542
                                                            Α5
                                                                       C
     116
          male
                70.5
                                         370369
                                                  7.7500
                                  0
                                                            NaN
     280
          male
                65.0
                                         336439
                                                   7.7500
                                                            NaN
                                                                       Q
     456
          male
                65.0
                                          13509
                                                 26.5500
                                                            E38
     493
                71.0
                                  0
                                       PC 17609
                                                 49.5042
                                                            NaN
          male
                                                 30.0000
     630
                80.0
                           0
                                  0
                                          27042
                                                            A23
                                                                       S
          male
     672
                70.0
                           0
                                  0
                                    C.A. 24580
                                                 10.5000
                                                            NaN
                                                                       S
          male
                                      WE/P 5735
                                                 71.0000
     745
          male
                70.0
                          1
                                  1
                                                            B22
                                                                       S
     851
          male
                74.0
                                         347060
                                                  7.7750
                                                            NaN
print(df[(df["Age"] < low) | (df["Age"] > up)].any(axis=None)) #True
Tindex([33, 54, 96, 116, 280, 456, 493, 630, 672, 745, 851], dtype='int64')
     False
def outlier_thresholds(dataframe, col_name, q1=0.25, q3=0.75):
    quartile1 = dataframe[col_name].quantile(q1)
    quartile3 = dataframe[col_name].quantile(q3)
    interquantile_range = quartile3 - quartile1
    up_limit = quartile3 + 1.5 * interquantile_range
    low_limit = quartile1 - 1.5 * interquantile_range
    return low_limit, up_limit
print(outlier_thresholds(df, "Age")) # (-6.6875, 64.8125)
low, up = outlier_thresholds(df, "Fare")
\label{eq:print}  \texttt{print}(\texttt{df}[\texttt{"Fare"}] < \texttt{low}) \ | \ (\texttt{df}[\texttt{"Fare"}] > \texttt{up})].\texttt{head}()) 
    (-6.6875, 64.8125)
         PassengerId
                      Survived Pclass
     27
                  28
                              0
                                      1
     31
                  32
                              1
                                      1
                  35
                              0
     34
                                      1
     52
                  53
                                      1
                              1
                                                        Name
                                                                 Sex
                                                                       Age
                                                                            SibSp
     1
         Cumings, Mrs. John Bradley (Florence Briggs Th...
                                                              female
                                                                      38.0
     27
                            Fortune, Mr. Charles Alexander
                                                                male
                                                                      19.0
                                                                                3
     31
            Spencer, Mrs. William Augustus (Marie Eugenie)
                                                                       NaN
     34
                                    Meyer, Mr. Edgar Joseph
                                                                male
                                                                      28.0
     52
                  Harper, Mrs. Henry Sleeper (Myna Haxtun)
                                                              female
                                                                      49.0
         Parch
                  Ticket
                               Fare
                                           Cabin Embarked
                            71,2833
     1
             0
                PC 17599
                                             C85
                                                        C
     27
                           263,0000
                                     C23 C25 C27
             2
                   19950
                                                        S
     31
                PC 17569
                           146.5208
                                             B78
                                                        C
     34
             0
                PC 17604
                            82,1708
                                             NaN
                                                        C
     52
             0 PC 17572
                            76.7292
                                             D33
                                                         C
def check_outlier(dataframe, col_name):
    low_limit, up_limit = outlier_thresholds(dataframe, col_name)
     if \ dataframe[(dataframe[col\_name] \ > \ up\_limit) \ | \ (dataframe[col\_name] \ < \ low\_limit)]. any(axis=None): \\
        return True
    else:
        return False
print(check_outlier(df, "Age"))
print(check_outlier(df, "Fare"))
    True
<del>_</del>
```

```
def grab_col_names(dataframe, cat_th=10, car_th=20):
    cat_cols = [col for col in dataframe.columns if dataframe[col].dtypes == "0"]
    num_but_cat = [col for col in dataframe.columns if dataframe[col].nunique() < cat_th and dataframe[col].dtypes != "0"]</pre>
    cat_but_car = [col for col in dataframe.columns if dataframe[col].nunique() > car_th and dataframe[col].dtypes == "0"]
    cat_cols = cat_cols + num_but_cat
    cat_cols = [col for col in cat_cols if col not in cat_but_car]
    num_cols = [col for col in dataframe.columns if dataframe[col].dtypes != "0" and col not in num_but_cat]
    print(f"Observations: {dataframe.shape[0]}")
    print(f"Variables: {dataframe.shape[1]}")
    print(f"cat_cols: {len(cat_cols)}")
    print(f"num_cols: {len(num_cols)}")
    print(f"cat_but_car: {len(cat_but_car)}")
    print(f"num_but_cat: {len(num_but_cat)}")
    return cat_cols, num_cols, cat_but_car
cat_cols, num_cols, cat_but_car = grab_col_names(df)
num_cols = [col for col in num_cols if col not in "PassengerId"]
print(num_cols) # ['Age', 'Fare']
for col in num_cols:
    print(col, check_outlier(df, col))
→ Observations: 891
     Variables: 12
     cat_cols: 6
     num_cols: 3
     cat_but_car: 3
     num_but_cat: 4
     ['Age', 'Fare']
     Age True
     Fare True
df = load()
low, up = outlier_thresholds(df, "Fare")
print(df.shape) # (891, 12)
\label{eq:print}  \texttt{print}(\texttt{df}[\sim((\texttt{df}["Fare"] < low) \ | \ (\texttt{df}["Fare"] > up))].shape) \ \#(775,12) 
def remove_outlier(dataframe, col_name):
    low_limit, up_limit = outlier_thresholds(dataframe, col_name)
    df_without_outliers = dataframe[~((dataframe[col_name] < low_limit) | (dataframe[col_name] > up_limit)))]
    return df_without_outliers
cat_cols, num_cols, cat_but_car = grab_col_names(df)
→ (891, 12)
     (775, 12)
     Observations: 891
     Variables: 12
     cat_cols: 6
     num_cols: 3
     cat_but_car: 3
     num_but_cat: 4
num_cols.remove('PassengerId')
for col in num_cols:
    df = remove_outlier(df,col)
print(df.shape) # (765,12)
→ (765, 12)
def replace_with_thresholds(dataframe, variable):
    low_limit, up_limit = outlier_thresholds(dataframe, variable)
    dataframe.loc[(dataframe[variable] < low_limit), variable] = low_limit</pre>
    dataframe.loc[(dataframe[variable] > up_limit), variable] = up_limit
df = load()
cat_cols, num_cols, cat_but_car = grab_col_names(df)
→ Observations: 891
     Variables: 12
     cat_cols: 6
     num_cols: 3
     cat_but_car: 3
     num_but_cat: 4
```

```
num_cols.remove('PassengerId')
for col in num_cols:
   print(col, check_outlier(df, col))
   Age True
    Fare True
for col in num_cols:
   replace_with_thresholds(df, col)
for col in num_cols:
   print(col, check_outlier(df, col))
    Age False
    Fare False
df = sns.load_dataset('diamonds')
print(df.shape) # (53940, 10)
print(df.head())
→ (53940, 10)
                  cut color clarity depth table price
       carat
    0
                       Е
                                                    326 3.95 3.98 2.43
                Ideal
        0.23
                               SI2 61.5
                                            55.0
    1
        0.21 Premium
                         Е
                               SI1
                                     59.8
                                            61.0
                                                    326 3.89 3.84 2.31
        0.23
                 Good
                         Ε
                               VS1 56.9
                                            65.0
                                                    327 4.05 4.07 2.31
    3
        0.29 Premium
                         Ι
                               VS2
                                     62.4
                                            58.0
                                                    334 4.20 4.23
                                                                    2.63
                             SI2 63.3 58.0 335 4.34 4.35 2.75
        0.31
df = df.select_dtypes(include=['float64', 'int64'])
df = df.dropna()
print(df.shape)
                 # (53940, 7)
print(df.head())
→ (53940, 7)
       carat depth table price
    0
        0.23
               61.5
                      55.0
                            326 3.95 3.98 2.43
        0.21
               59.8
                      61.0
                              326 3.89 3.84 2.31
        0.23
               56.9
                      65.0
                             327 4.05 4.07 2.31
               62.4
                      58.0
                              334 4.20 4.23 2.63
        0.29
        0.31
              63.3
                     58.0
                             335 4.34 4.35 2.75
for col in df.columns:
   print(col, check_outlier(df, col))
→ carat True
    depth True
    table True
    price True
    x True
    y True
    z True
low, up = outlier_thresholds(df, "carat")
print(df[((df["carat"] < low) | (df["carat"] > up))].shape) # (1889, 7)
low, up = outlier_thresholds(df, "depth")
\label{lower_print}  print(df[((df["depth"] < low) \ | \ (df["depth"] > up))].shape) \ \# \ (2545, \ 7) 
clf = LocalOutlierFactor(n_neighbors=20)
clf.fit_predict(df)
df_scores = clf.negative_outlier_factor_
print(df_scores)
print(np.sort(df_scores)[0:5])
scores = pd.DataFrame(np.sort(df_scores))
scores.plot(stacked=True, xlim=[0, 20], style='.-')
plt.show()
```

```
(1889, 7)
(2545, 7)
[-1.58352526 -1.59732899 -1.62278873 ... -1.06721815 -1.00809552
-1.00849038]
[-8.60843658 -8.20889984 -5.86084355 -4.98415175 -4.81502092]
```

