- 1 from google.colab import drive
- 1 drive.mount('/content/gdrive')

Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive.mount("/content/gdrive", force_remount=True).

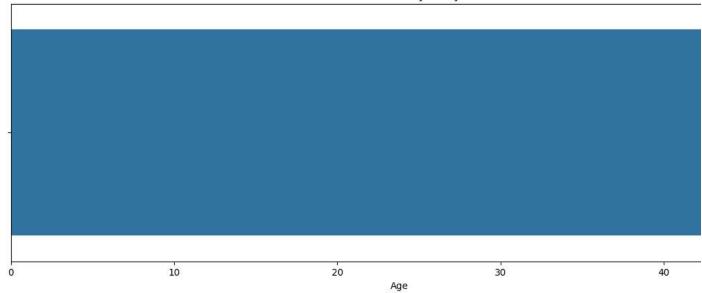
- 1 import pandas as pd
- 2 import numpy as np
- 3 import os
- 4 import matplotlib.pyplot as plt
- 5 from sklearn.decomposition import PCA
- 1 df = pd.read_csv('/content/gdrive/MyDrive/mcdonalds.csv')
- 1 df.head()

	yummy	convenient	spicy	fattening	greasy	fast	cheap	tasty	expensive	healthy	disgusting	Like	Age	VisitFrequency	Gender
0	No	Yes	No	Yes	No	Yes	Yes	No	Yes	No	No	-3	61	Every three months	Female
1	Yes	Yes	No	Yes	Yes	Yes	Yes	Yes	Yes	No	No	+2	51	Every three months	Female
2	No	Yes	Yes	Yes	Yes	Yes	No	Yes	Yes	Yes	No	+1	62	Every three months	Female
3	Yes	Yes	No	Yes	Yes	Yes	Yes	Yes	No	No	Yes	+4	69	Once a week	Female
4	No	Yes	No	Yes	Yes	Yes	Yes	No	No	Yes	No	+2	49	Once a month	Male

- 1 import seaborn as sns
- 2 plt.figure(figsize=(15,5))
- 3 plt.title('distribution of yummy')
- 4 sns.barplot(x='Age',data=df)

<Axes: title={'center': 'distribution of yummy'}, xlabel='Age'>

distribution of yummy



1 df.describe(include="all")

```
yummy convenient spicy fattening greasy fast cheap tasty expensive healthy disgusting Like
                                                                                                                            Age VisitFrequency
     count
              1453
                          1453
                                 1453
                                            1453
                                                    1453
                                                          1453
                                                                 1453
                                                                        1453
                                                                                   1453
                                                                                            1453
                                                                                                        1453
                                                                                                              1453 1453.000000
                                                                                                                                            1453
                 2
                             2
                                    2
                                               2
                                                       2
                                                             2
                                                                    2
                                                                           2
                                                                                      2
                                                                                               2
                                                                                                           2
                                                                                                                                              6
     unique
                                                                                                                11
                                                                                                                           NaN
               Yes
                           Yes
                                   No
                                             Yes
                                                           Yes
                                                                         Yes
                                                                                              No
                                                                                                          No
                                                                                                                +3
                                                                                                                            NaN
       top
                                                     Yes
                                                                  Yes
                                                                                     No
                                                                                                                                    Once a month
      freq
               803
                          1319
                                 1317
                                            1260
                                                     765 1308
                                                                  870
                                                                         936
                                                                                    933
                                                                                            1164
                                                                                                         1100
                                                                                                               229
                                                                                                                            NaN
                                                                                                                                            439
1 df.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 1453 entries, 0 to 1452
    Data columns (total 15 columns):
         Column
                         Non-Null Count Dtype
    ---
                          -----
     0
         yummy
                          1453 non-null
                                          object
                          1453 non-null
     1
         convenient
                                          object
     2
         spicv
                          1453 non-null
                                          object
     3
         fattening
                          1453 non-null
                                          object
     4
                          1453 non-null
                                          object
         greasy
     5
         fast
                          1453 non-null
                                          object
                          1453 non-null
     6
         cheap
                                          object
     7
         tasty
                          1453 non-null
                                          object
     8
         expensive
                          1453 non-null
                                          object
         healthy
                          1453 non-null
                                          object
     10
         disgusting
                          1453 non-null
                                          object
     11
         Like
                          1453 non-null
                                          object
                          1453 non-null
                                          int64
     12
         Age
         VisitFrequency 1453 non-null
     13
                                          object
         Gender
                          1453 non-null
                                          object
    dtypes: int64(1), object(14)
    memory usage: 170.4+ KB
1 df.shape
    (1453, 15)
1 #check NaN Values
2 df.isna().sum()
    yummy
    convenient
                      0
                      0
    spicy
    fattening
                      0
    greasy
                      0
    fast
                      0
    cheap
                      0
    tasty
                      0
    expensive
                      0
    healthy
                      0
    disgusting
                      0
    Like
                      0
    Age
    VisitFrequency
                      0
    Gender
    dtype: int64
1 df.Age.describe([.75,.90,.95,.99])
             1453.000000

→ count

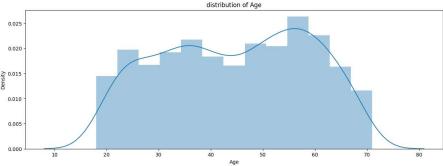
    mean
               44.604955
               14.221178
    std
               18.000000
    min
               45.000000
    50%
    75%
               57.000000
    90%
               63.000000
    95%
               66.000000
    99%
               70.000000
               71.000000
    max
    Name: Age, dtype: float64
1 #distribution of Age
2 plt.figure(figsize=(15,5))
3 plt.title("distribution of Age")
4 sns.distplot(df['Age'])
```

```
<ipython-input-19-34b028f839f5>:4: UserWarning:
```

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751



```
1 from sklearn.preprocessing import LabelEncoder
 2 le = LabelEncoder()
 3 df["yummy"]=le.fit transform(df["yummy"])
 4 df["convenient"]=le.fit_transform(df["convenient"])
 5 df["spicy"]=le.fit_transform(df["spicy"])
 6 df["fattening"]=le.fit_transform(df["fattening"])
 7 df["greasy"]=le.fit_transform(df["greasy"])
 8 df["fast"]=le.fit transform(df["fast"])
 9 df["cheap"]=le.fit_transform(df["cheap"])
10 df["tasty"]=le.fit_transform(df["tasty"])
11 df["expensive"]=le.fit_transform(df["expensive"])
12 df["healthy"]=le.fit_transform(df["healthy"])
13 df["disgusting"]=le.fit_transform(df["disgusting"])
14 df["Like"]=le.fit_transform(df["Like"])
15 df["Age"]=le.fit_transform(df["Age"])
16 df["VisitFrequency"]=le.fit_transform(df["VisitFrequency"])
17 df["Gender"]=le.fit_transform(df["Gender"])
1 df.shape
    (1453, 15)
 1 df.columns
    'VisitFrequency', 'Gender'],
          dtype='object')
 1 # columns to keep:
 2 data= df[['yummy', 'convenient', 'spicy', 'fattening', 'greasy', 'fast', 'cheap',
         'tasty', 'expensive', 'healthy', 'disgusting', 'Like', 'Age',
         'VisitFrequency', 'Gender']].rename({'Gender':'label'},axis=1)
 1 df.head(2)
```

```
1 from sklearn.model_selection import train_test_split
                      7
                             U
                                        1
                                               U
1 X = data.iloc[:, data.columns != 'label']
2 y = data.iloc[:, data.columns == 'label']
1 # split the data into test and train by maintaing same distribution of output variable 'y_true'[stratify=y_true]
2 X_train, test_df, y_train, y_test = train_test_split(X, y, stratify=y,test_size=0.2)
3 # split the train data into train and cross calidation by maintaining same distrubution of output varaible 'y_train'[stratify=y_true]
4 train_df, cv_df, ytrain, y_cv = train_test_split(X_train, y_train,stratify=y_train, test_size=0.2)
1 print('Number of data points in train data:', train_df.shape[0])
2 print('Number of data points in test data:', test_df.shape[0])
3 print('Number of data points in cross validation data:',df.shape[0])
   Number of data points in train data: 929
   Number of data points in test data: 291
   Number of data points in cross validation data: 1453
1 test df.head(2)
          yummy convenient spicy fattening greasy fast cheap tasty expensive health;
                                                                                   0
    1090
              n
                          1
                                 0
                                            1
                                                    0
                                                          n
                                                                 0
    1252
1 y_test.head(2)
          label
    1090
              1
    1252
              1
1 from sklearn.neighbors import KNeighborsClassifier
1 knn = KNeighborsClassifier()
1 knn.fit(X_train,y_train)
   /usr/local/lib/python3.10/dist-packages/sklearn/neighbors/_classification.py:215: DataCo
     return self._fit(X, y)
    ▼ KNeighborsClassifier
    KNeighborsClassifier()
1 kYPred = knn.predict(X train)
1 from sklearn.metrics import classification_report
2 from sklearn.metrics import confusion matrix
1 print(classification_report(y_train,kYPred))
                  precision
                               recall f1-score
                                                  support
              0
                       0.69
                                 9.78
                                           0.73
                                                      630
                       0.69
                                           0.63
                                                      532
                                           0.69
       accuracy
                                                     1162
       macro avg
                       0.69
                                 0.68
                                           0.68
                                                     1162
   weighted avg
                       0.69
                                 0.69
                                           0.69
                                                     1162
1 confusion_matrix(y_train,kYPred)
```

array([[489, 141], [221, 311]])

1 import pickle

1 pickle.dump(knn,open('knnmodel','wb'))