In [1]: import pandas as pd
 import numpy as np
 import matplotlib.pyplot as plt
 import seaborn as sns

C:\ProgramData\Anaconda3\lib\importlib_bootstrap.py:219: RuntimeWarning: nump y.ufunc size changed, may indicate binary incompatibility. Expected 192 from C header, got 216 from PyObject

return f(*args, **kwds)

C:\ProgramData\Anaconda3\lib\importlib_bootstrap.py:219: RuntimeWarning: nump y.ufunc size changed, may indicate binary incompatibility. Expected 192 from C header, got 216 from PyObject

return f(*args, **kwds)

C:\ProgramData\Anaconda3\lib\importlib_bootstrap.py:219: RuntimeWarning: nump y.ufunc size changed, may indicate binary incompatibility. Expected 192 from C header, got 216 from PyObject

return f(*args, **kwds)

C:\ProgramData\Anaconda3\lib\importlib_bootstrap.py:219: RuntimeWarning: nump
y.ufunc size changed, may indicate binary incompatibility. Expected 216, got 19
2

return f(*args, **kwds)

C:\ProgramData\Anaconda3\lib\importlib_bootstrap.py:219: RuntimeWarning: nump y.ufunc size changed, may indicate binary incompatibility. Expected 192 from C header, got 216 from PyObject

return f(*args, **kwds)

In [2]: df=pd.read_csv("C:/Users/DELL/Desktop/pizza.csv")

In [3]: | df.head()

Out[3]:

	id	dateAdded	dateUpdated	address	categories	primaı
0	AVz3Y-7h3D1zeR_xDAqm	2017-06- 30T05:05:40Z	2019-05- 01T15:43:09Z	4203 E Kiehl Ave	Pizza,Restaurant,American restaurants,Pizza Pl	Acco F
1	AVweGPFF_7pvs4fzAAzQ	2016-04- 02T04:02:49Z	2019-05- 01T15:27:50Z	25 E Camelback Rd	Pizza,Pizza Place,Restaurants	Acco F
2	AVwdRGa9_7pvs4fz4E3K	2016-03- 03T18:39:49Z	2019-05- 01T12:52:25Z	3703 Paxton Ave	Restaurant,Pizza Place,Restaurants	Acco F
3	AVwdX4psIN2L1WUfvJB1	2016-03- 29T05:08:59Z	2019-05- 01T12:52:20Z	30495 John R Rd	Pizza,Carry-out food,Pizza Place,Restaurants	Acco F
4	AVwdaeTtkufWRAb55pSH	2016-03- 31T02:34:04Z	2019-05- 01T12:50:45Z	3600 Eastern Ave	Pizza,American restaurants,Pizza Place,Pizza e	Acco F

5 rows × 24 columns

<class 'pandas.core.frame.DataFrame'>

```
In [4]: df.info()
```

```
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 24 columns):
id
                      10000 non-null object
dateAdded
                      10000 non-null object
dateUpdated
                      10000 non-null object
                      10000 non-null object
address
categories
                      10000 non-null object
                      10000 non-null object
primaryCategories
                      10000 non-null object
city
                      10000 non-null object
country
keys
                      10000 non-null object
latitude
                      10000 non-null float64
longitude
                      10000 non-null float64
                      1679 non-null object
menuPageURL
                      10000 non-null float64
menus.amountMax
                      10000 non-null float64
menus.amountMin
                      10000 non-null object
menus.currency
menus.dateSeen
                      10000 non-null object
                      3718 non-null object
menus
                      10000 non-null object
menus.name
name
                      10000 non-null object
                      9996 non-null object
postalCode
priceRangeCurrency
                      10000 non-null object
priceRangeMin
                      10000 non-null int64
priceRangeMax
                      10000 non-null int64
province
                      10000 non-null object
dtypes: float64(4), int64(2), object(18)
memory usage: 1.8+ MB
```

In [7]: df.shape

Out[7]: (10000, 24)

In [8]: df.describe()

Out[8]:

	latitude	longitude	menus.amountMax	menus.amountMin	priceRangeMin	priceRan
count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.0
mean	38.371895	-90.063800	14.032670	13.416902	4.655500	27.7
std	4.573537	17.340379	17.607233	10.742678	9.828412	6.0
min	21.421326	-157.802443	0.000000	0.000000	0.000000	7.0
25%	34.422946	-104.801126	8.490000	8.000000	0.000000	25.0
50%	40.118880	-82.909949	13.390000	12.990000	0.000000	25.0
75%	40.905048	-75.194741	17.950000	16.990000	0.000000	25.0
max	64.850500	-71.946170	1395.000000	243.000000	50.000000	55.0
4						

In [9]: df_cat=df.select_dtypes(include=[object])
 df_num=df.select_dtypes(include=[np.number])

In [10]: df_cat.head()

Out[10]:

	id	dateAdded	dateUpdated	address	categories	primaı
0	AVz3Y-7h3D1zeR_xDAqm	2017-06- 30T05:05:40Z	2019-05- 01T15:43:09Z	4203 E Kiehl Ave	Pizza,Restaurant,American restaurants,Pizza Pl	Acco F
1	AVweGPFF_7pvs4fzAAzQ	2016-04- 02T04:02:49Z	2019-05- 01T15:27:50Z	25 E Camelback Rd	Pizza,Pizza Place,Restaurants	Acco F
2	AVwdRGa9_7pvs4fz4E3K	2016-03- 03T18:39:49Z	2019-05- 01T12:52:25Z	3703 Paxton Ave	Restaurant,Pizza Place,Restaurants	Acco F
3	AVwdX4psIN2L1WUfvJB1	2016-03- 29T05:08:59Z	2019-05- 01T12:52:20Z	30495 John R Rd	Pizza,Carry-out food,Pizza Place,Restaurants	Acco F
4	AVwdaeTtkufWRAb55pSH	2016-03- 31T02:34:04Z	2019-05- 01T12:50:45Z	3600 Eastern Ave	Pizza,American restaurants,Pizza Place,Pizza e	Acco F
4						

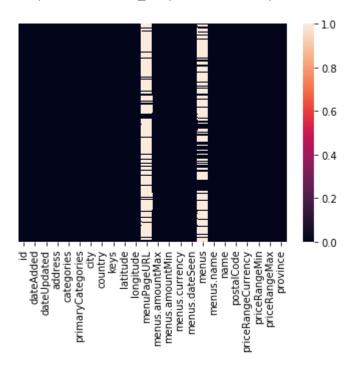
In [11]: df_num.head()

Out[11]:

	latitude	longitude	menus.amountMax	menus.amountMin	priceRangeMin	priceRangeMax
0	34.832300	-92.183800	7.98	7.98	0	25
1	33.509266	-112.073044	6.00	6.00	0	25
2	39.144883	-84.432685	6.49	6.49	0	25
3	42.516669	-83.106630	5.99	5.99	25	40
4	39.286630	-76.566984	5.49	5.49	0	25

In [12]: sns.heatmap(df.isnull(),yticklabels=False)

Out[12]: <matplotlib.axes._subplots.AxesSubplot at 0x24ab4118320>



In [13]: df_cat.isnull()

Out[13]:

	id	dateAdded	dateUpdated	address	categories	primaryCategories	city	country	key
0	False	False	False	False	False	False	False	False	Fals
1	False	False	False	False	False	False	False	False	Fals
2	False	False	False	False	False	False	False	False	Fals
3	False	False	False	False	False	False	False	False	Fals
4	False	False	False	False	False	False	False	False	Fals
5	False	False	False	False	False	False	False	False	Fals
6	False	False	False	False	False	False	False	False	Fals
7	False	False	False	False	False	False	False	False	Fals
8	False	False	False	False	False	False	False	False	Fals
9	False	False	False	False	False	False	False	False	Fals
10	False	False	False	False	False	False	False	False	Fals
11	False	False	False	False	False	False	False	False	Fals
12	False	False	False	False	False	False	False	False	Fals
13	False	False	False	False	False	False	False	False	Fals
14	False	False	False	False	False	False	False	False	Fals
15	False	False	False	False	False	False	False	False	Fals
16	False	False	False	False	False	False	False	False	Fals
17	False	False	False	False	False	False	False	False	Fals
18	False	False	False	False	False	False	False	False	Fals
19	False	False	False	False	False	False	False	False	Fals
20	False	False	False	False	False	False	False	False	Fals
21	False	False	False	False	False	False	False	False	Fals
22	False	False	False	False	False	False	False	False	Fals
23	False	False	False	False	False	False	False	False	Fals
24	False	False	False	False	False	False	False	False	Fals
25	False	False	False	False	False	False	False	False	Fals
26	False	False	False	False	False	False	False	False	Fals
27	False	False	False	False	False	False	False	False	Fals
28	False	False	False	False	False	False	False	False	Fals
29	False	False	False	False	False	False	False	False	Fals
									•
9970	False	False	False	False	False	False	False	False	Fals
9971	False	False	False	False	False	False	False	False	Fals
9972	False	False	False	False	False	False	False	False	Fals

	id	dateAdded	dateUpdated	address	categories	primaryCategories	city	country	key
9973	False	False	False	False	False	False	False	False	Fals
9974	False	False	False	False	False	False	False	False	Fals
9975	False	False	False	False	False	False	False	False	Fals
9976	False	False	False	False	False	False	False	False	Fals
9977	False	False	False	False	False	False	False	False	Fals
9978	False	False	False	False	False	False	False	False	Fals
9979	False	False	False	False	False	False	False	False	Fals
9980	False	False	False	False	False	False	False	False	Fals
9981	False	False	False	False	False	False	False	False	Fals
9982	False	False	False	False	False	False	False	False	Fals
9983	False	False	False	False	False	False	False	False	Fals
9984	False	False	False	False	False	False	False	False	Fals
9985	False	False	False	False	False	False	False	False	Fals
9986	False	False	False	False	False	False	False	False	Fals
9987	False	False	False	False	False	False	False	False	Fals
9988	False	False	False	False	False	False	False	False	Fals
9989	False	False	False	False	False	False	False	False	Fals
9990	False	False	False	False	False	False	False	False	Fals
9991	False	False	False	False	False	False	False	False	Fals
9992	False	False	False	False	False	False	False	False	Fals
9993	False	False	False	False	False	False	False	False	Fals
9994	False	False	False	False	False	False	False	False	Fals
9995	False	False	False	False	False	False	False	False	Fals
9996	False	False	False	False	False	False	False	False	Fals
9997	False	False	False	False	False	False	False	False	Fals
9998	False	False	False	False	False	False	False	False	Fals
9999	False	False	False	False	False	False	False	False	Fals

10000 rows × 18 columns

```
In [14]: df cat.isnull().sum()
Out[14]: id
                                     0
          dateAdded
                                     0
          dateUpdated
                                     0
          address
                                     0
                                     0
          categories
          primaryCategories
                                     0
                                     0
          city
          country
                                     0
                                     0
          keys
                                  8321
          menuPageURL
          menus.currency
                                     0
          menus.dateSeen
                                     0
                                  6282
          menus
          menus.name
                                     0
                                     0
          name
                                     4
          postalCode
                                     0
          priceRangeCurrency
          province
                                     0
          dtype: int64
```

In [29]: df_cat.menuPageURL.fillna(df_cat.menuPageURL.value_counts().idxmax(),inplace=True
df_cat.menus.fillna(df_cat.menus.value_counts().idxmax(),inplace=True)
df_cat.postalCode.fillna(df_cat.postalCode.value_counts().idxmax(),inplace=True)

C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\generic.py:6130: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stab le/indexing.html#indexing-view-versus-copy (http://pandas.pydata.org/pandas-doc s/stable/indexing.html#indexing-view-versus-copy)

self._update_inplace(new_data)

In [30]: | df_cat.head()

Out[30]:

	id	dateAdded	dateUpdated	address	categories	primaı
0	AVz3Y-7h3D1zeR_xDAqm	2017-06- 30T05:05:40Z	2019-05- 01T15:43:09Z	4203 E Kiehl Ave	Pizza,Restaurant,American restaurants,Pizza Pl	Acco F
1	AVweGPFF_7pvs4fzAAzQ	2016-04- 02T04:02:49Z	2019-05- 01T15:27:50Z	25 E Camelback Rd	Pizza,Pizza Place,Restaurants	Acco F
2	AVwdRGa9_7pvs4fz4E3K	2016-03- 03T18:39:49Z	2019-05- 01T12:52:25Z	3703 Paxton Ave	Restaurant,Pizza Place,Restaurants	Acco F
3	AVwdX4psIN2L1WUfvJB1	2016-03- 29T05:08:59Z	2019-05- 01T12:52:20Z	30495 John R Rd	Pizza,Carry-out food,Pizza Place,Restaurants	Acco F
4	AVwdaeTtkufWRAb55pSH	2016-03- 31T02:34:04Z	2019-05- 01T12:50:45Z	3600 Eastern Ave	Pizza,American restaurants,Pizza Place,Pizza e	Acco F

In [31]: df_num.isnull().sum()

Out[31]: latitude 0
longitude 0
menus.amountMax 0
menus.amountMin 0
priceRangeMin 0
priceRangeMax 0
dtype: int64

In [32]: df_num.head()

Out[32]:

	latitude	longitude	menus.amountMax	menus.amountMin	priceRangeMin	priceRangeMax
C	34.832300	-92.183800	7.98	7.98	0	25
1	33.509266	-112.073044	6.00	6.00	0	25
2	39.144883	-84.432685	6.49	6.49	0	25
3	42.516669	-83.106630	5.99	5.99	25	40
4	39.286630	-76.566984	5.49	5.49	0	25

In [33]: from sklearn.preprocessing import LabelEncoder

In [34]: le=LabelEncoder()

```
df_cat=df_cat.apply(le.fit_transform)
In [35]:
              df=pd.concat([df_cat,df_num],axis=1)
In [37]:
              df.head()
In [38]:
Out[38]:
                                   name postalCode priceRangeCurrency
                                                                                                      latitude
                                                                                                                    longitude menus.an
             nuPageURL
                         14
                                                    1222
                                                                                  0
                                    1540
                                                                                                   34.832300
                                                                                                                   -92.183800
                                                                                               2
                         10
                                    1519
                                                    1409
                                                                                  0
                                                                                               3
                                                                                                   33.509266
                                                                                                                 -112.073044
                        119
                                    1011
                                                     873
                                                                                   0
                                                                                              31
                                                                                                   39.144883
                                                                                                                  -84.432685
                                                     949
                                                                                                                  -83.106630
                         10
                                     684
                                                                                              19
                                                                                                   42.516669
                         10
                                    1581
                                                     418
                                                                                              18
                                                                                                   39.286630
                                                                                                                  -76.566984
 In [ ]:
 In [ ]:
               sns.heatmap(df.isnull(),yticklabels=False)
In [39]:
Out[39]: <matplotlib.axes._subplots.AxesSubplot at 0x24ab8cf5710>
                                                                             - 0.08
                                                                               0.04
                                                                              - 0.00
                                                                               -0.04
                   dateAdded -
dateUpdated -
address -
categories -
primaryCategories -
                                                   postalCode -
priceRangeCurrency -
province -
latitude -
longitude -
                                                                menus.amountMin -
priceRangeMin -
priceRangeMax -
                                              menus.name
name
                                country
                                                              menus.amountMax
                                          menus.dateSeel
```

In [79]: df.head()

Out[79]:

long	latitude	province	priceRangeCurrency	postalCode	name	menus.name	 enus.dateSeen
-92.18	34.832300	2	0	1222	1540	1106	 5492
-112.07	33.509266	3	0	1409	1519	3158	 4602
-84.43	39.144883	31	0	873	1011	3110	 5173
-83.10	42.516669	19	0	949	684	3919	 634
-76.56	39.286630	18	0	418	1581	3382	 278

In [81]: df.drop(["priceRangeCurrency"],axis=1,inplace=True)

In [82]: df.head()

Out[82]:

menuPageURL	menus.currency	menus.dateSeen	menus	menus.name	name	postalCode	province
14	0	5492	2707	1106	1540	1222	2
10	0	4602	2707	3158	1519	1409	3
119	0	5173	2792	3110	1011	873	31
10	0	634	2707	3919	684	949	19
10	0	278	2707	3382	1581	418	18
•							>

In [87]: df.drop(["id","dateAdded","dateUpdated","latitude"],axis=1,inplace=True)

In [88]: df.head()

Out[88]:

geURL	menus.currency	menus.dateSeen	menus	menus.name	name	postalCode	province	long
14	0	5492	2707	1106	1540	1222	2	-92.18
10	0	4602	2707	3158	1519	1409	3	-112.07
119	0	5173	2792	3110	1011	873	31	-84.43
10	0	634	2707	3919	684	949	19	-83.10
10	0	278	2707	3382	1581	418	18	-76.56
•								•

In [89]: df.shape

Out[89]: (10000, 16)

```
from sklearn.model selection import KFold
In [104]: kf=KFold(n_splits=10)
In [123]: | x=df.iloc[:,[0,1,2,3,4,5,6,7,8,9,10,11,12,13,14]].values
          y=df.iloc[:,15].values
In [109]: for a,b in kf.split(x):
              x_train,x_test=x[a],x[b]
              y_train,y_test=y[a],y[b]
              x1_train,x1_test=x[a],x[b]
              y1_train,y1_test=y[a],y[b]
              x2_train,x2_test=x[a],x[b]
              y2_train,y2_test=y[a],y[b]
              x3_train,x3_test=x[a],x[b]
              y3_train,y3_test=y[a],y[b]
              x4_train,x4_test=x[a],x[b]
              y4_train,y4_test=y[a],y[b]
              x5_train,x5_test=x[a],x[b]
              y5_train,y5_test=y[a],y[b]
In [110]:
          from sklearn.tree import DecisionTreeClassifier
          from sklearn.neighbors import KNeighborsClassifier
          from sklearn.naive_bayes import GaussianNB
           from sklearn.svm import LinearSVC
           from sklearn.ensemble import RandomForestClassifier
           from sklearn.linear_model import LogisticRegression
In [111]:
          dt=DecisionTreeClassifier()
          knn=KNeighborsClassifier()
          nb=GaussianNB()
           svm=LinearSVC()
          rf=RandomForestClassifier()
          lr=LogisticRegression()
```

```
In [112]:
          dt.fit(x train,y train)
          knn.fit(x1 train,y1 train)
          nb.fit(x2 train,y2 train)
          svm.fit(x3 train,y3 train)
          rf.fit(x4_train,y4_train)
          lr.fit(x5_train,y5_train)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\svm\base.py:931: Convergence
          Warning: Liblinear failed to converge, increase the number of iterations.
            "the number of iterations.", ConvergenceWarning)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\ensemble\forest.py:246: Futu
          reWarning: The default value of n_estimators will change from 10 in version 0.2
          0 to 100 in 0.22.
            "10 in version 0.20 to 100 in 0.22.", FutureWarning)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:43
          3: FutureWarning: Default solver will be changed to 'lbfgs' in 0.22. Specify a
          solver to silence this warning.
            FutureWarning)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear model\logistic.py:46
          0: FutureWarning: Default multi class will be changed to 'auto' in 0.22. Specif
          y the multi class option to silence this warning.
            "this warning.", FutureWarning)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\svm\base.py:931: Convergence
          Warning: Liblinear failed to converge, increase the number of iterations.
            "the number of iterations.", ConvergenceWarning)
Out[112]: LogisticRegression(C=1.0, class weight=None, dual=False, fit intercept=True,
                    intercept scaling=1, max iter=100, multi class='warn',
                    n_jobs=None, penalty='12', random_state=None, solver='warn',
                    tol=0.0001, verbose=0, warm_start=False)
In [113]:
          pred=dt.predict(x_test)
          pred1=knn.predict(x1 test)
          pred2=nb.predict(x2 test)
          pred3=svm.predict(x3 test)
          pred4=rf.predict(x4_test)
          pred5=lr.predict(x5 test)
In [117]: from sklearn.metrics import confusion matrix
In [118]:
          cm=confusion matrix(y test,pred)
          cm1=confusion_matrix(y1_test,pred1)
          cm2=confusion matrix(y2 test,pred2)
          cm3=confusion_matrix(y3_test,pred3)
          cm4=confusion_matrix(y4_test,pred4)
           cm5=confusion matrix(y5 test,pred5)
```

```
In [119]: cm, cm1, cm2, cm3, cm4, cm5
Out[119]: (array([[684, 134,
                                 0],
                   [122,
                          56,
                                0],
                      3,
                           1,
                                 0]], dtype=int64), array([[765, 52,
                                                                         1],
                   [158,
                          20,
                                 0],
                                 0]], dtype=int64), array([[660, 144,
                                                                        14,
                                                                               0],
                      4,
                     0,
                           0,
                                 0,
                                      0],
                                      0],
                   [144,
                          31,
                                 3,
                   [ 4,
                                      0]], dtype=int64), array([[818,
                           0,
                                 0,
                                                                               0],
                   [178,
                           0,
                                 0],
                                 0]], dtype=int64), array([[800,
                                                                         0],
                   [ 4,
                           0,
                   [147,
                          31,
                                 0],
                                 0]], dtype=int64), array([[ 0,
                           2,
                                                                    0,
                                                                               0],
                      2,
                                                                         0,
                      0, 818,
                                      0],
                                 0,
                      1, 177,
                                 0,
                                      0],
                                      0]], dtype=int64))
                      0,
                           4,
                                 0,
In [120]: from sklearn.metrics import accuracy score
In [121]: | a=accuracy_score(y_test,pred)
           b=accuracy_score(y1_test,pred1)
           c=accuracy_score(y2_test,pred2)
           d=accuracy_score(y3_test,pred3)
           e=accuracy_score(y4_test,pred4)
           f=accuracy_score(y5_test,pred5)
In [122]: a,b,c,d,e,f
Out[122]: (0.74, 0.785, 0.663, 0.818, 0.831, 0.818)
  In [ ]:
```