```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

import seaborn as sns

from sklearn.linear_model import LinearRegression from sklearn.model_selection import train_test_split

from sklearn.metrics import r2_score

df = pd.read_csv(r'/content/avocado-csv.csv')

df

	Unnamed: 0	Date	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags	Larg Bag
0	0	12/27/2015	1.33	64236.62	1036.74	54454.85	48.16	8696.87	8603.62	93.2
1	1	12/20/2015	1.35	54876.98	674.28	44638.81	58.33	9505.56	9408.07	97.4
2	2	12/13/2015	0.93	118220.22	794.70	109149.67	130.50	8145.35	8042.21	103.1
3	3	12/6/2015	1.08	78992.15	1132.00	71976.41	72.58	5811.16	5677.40	133.7
4	4	11/29/2015	1.28	51039.60	941.48	43838.39	75.78	6183.95	5986.26	197.6
18244	7	2/4/2018	1.63	17074.83	2046.96	1529.20	0.00	13498.67	13066.82	431.8
18245	8	1/28/2018	1.71	13888.04	1191.70	3431.50	0.00	9264.84	8940.04	324.8
18246	9	1/21/2018	1.87	13766.76	1191.92	2452.79	727.94	9394.11	9351.80	42.3
18247	10	1/14/2018	1.93	16205.22	1527.63	2981.04	727.01	10969.54	10919.54	50.0
18248	11	1/7/2018	1.62	17489.58	2894.77	2356.13	224.53	12014.15	11988.14	26.0

18249 rows × 14 columns



df.head()

	Unnamed: 0	Date	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags	Large Bags	XLaı Ba
0	0	12/27/2015	1.33	64236.62	1036.74	54454.85	48.16	8696.87	8603.62	93.25	
1	1	12/20/2015	1.35	54876.98	674.28	44638.81	58.33	9505.56	9408.07	97.49	
2	2	12/13/2015	0.93	118220.22	794.70	109149.67	130.50	8145.35	8042.21	103.14	
3	3	12/6/2015	1.08	78992.15	1132.00	71976.41	72.58	5811.16	5677.40	133.76	
4	4	11/29/2015	1.28	51039.60	941.48	43838.39	75.78	6183.95	5986.26	197.69	



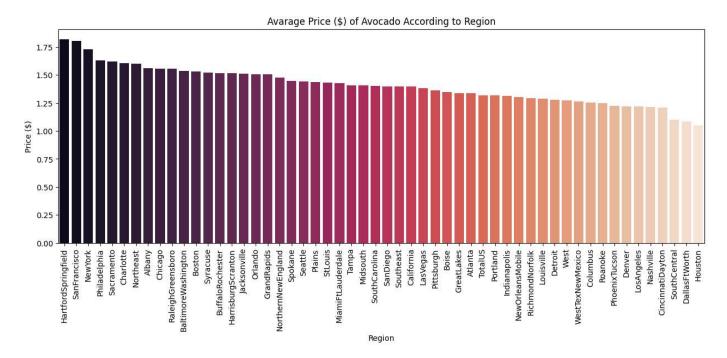
df = df.drop(columns=['Unnamed: 0','4046','4225','4770','Date'],axis=1)

df.head()



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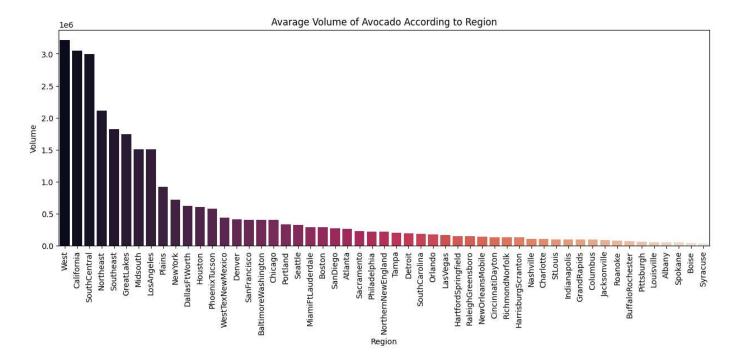
data1 = get_avarge_between_two_columns(df,'region','AveragePrice')
plot(data1,'Region','Price (\$)')



print(data1['column1'].iloc[-1], " is the region producing avocado with the lowest price.")
 Houston is the region producing avocado with the lowest price.

- $\ensuremath{\text{\#}}$ what is the highest region of avocado production
- # Checking if there are outlier values or not.

data2 = get_avarge_between_two_columns(df,'region','Total Volume')
sns.boxplot(x=data2.column2).set_title("Figure: Boxplot repersenting outlier columns.")



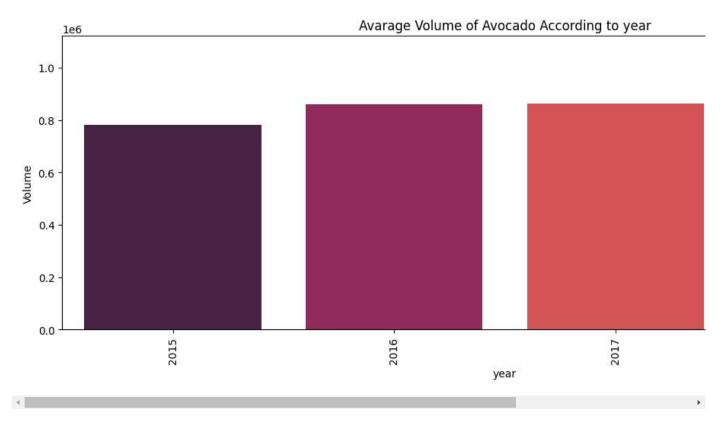
What is the average avocado prices in each year?

data3 = get_avarge_between_two_columns(df,'year','AveragePrice')
plot(data3,'year','Price')

```
Avarage Price of Avocado According to year
```

```
# What is the average avocado volume in each year?
```

```
data4 = get_avarge_between_two_columns(df,'year','Total Volume')
plot(data4,'year','Volume')
```



```
# Data Modeling
```

#we built the regrassion model by using linear regrassion from sklearn to predict the Avocado Price # Changing some column types in to catagories

```
df['region'] = df['region'].astype('category')
df['region'] = df['region'].cat.codes
df['type'] = df['type'].astype('category')
df['type'] = df['type'].cat.codes
```

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18249 entries, 0 to 18248
Data columns (total 9 columns):
Column Non-Null Count Dty

#	Column	Non-Null Count	Dtype				
0	AveragePrice	18249 non-null	float64				
1	Total Volume	18249 non-null	float64				
2	Total Bags	18249 non-null	float64				
3	Small Bags	18249 non-null	float64				
4	Large Bags	18249 non-null	float64				
5	XLarge Bags	18249 non-null	float64				
6	type	18249 non-null	int8				
7	year	18249 non-null	int64				
8	region	18249 non-null	int8				
dtypes: float64(6), int64(1), int8(2)							
memory usage: 1.0 MB							

df.head()

```
AveragePrice Total Volume Total Bags Small Bags Large Bags XLarge Bags type year region
      0
                 1.33
                           64236.62
                                         8696.87
                                                     8603.62
                                                                   93.25
                                                                                  0.0
                                                                                         0 2015
                                                                                                       0
      1
                 1.35
                           54876.98
                                         9505.56
                                                    9408.07
                                                                   97.49
                                                                                  0.0
                                                                                         0 2015
                                                                                                       0
      2
                 0.93
                          118220.22
                                        8145.35
                                                    8042.21
                                                                  103.14
                                                                                         0 2015
                                                                                  0.0
                                                                                                       0
      3
                 1.08
                           78992.15
                                         5811.16
                                                    5677.40
                                                                  133.76
                                                                                  0.0
                                                                                         0 2015
                                                                                                       0
      4
                 1.28
                           51039.60
                                        6183.95
                                                    5986.26
                                                                  197.69
                                                                                  0.0
                                                                                         0 2015
                                                                                                       0
# split data into x and y
X = df.drop(['AveragePrice'],axis=1)
y = df['AveragePrice']
# split data into training and testing dataset
X_train,X_test, y_train, y_test = train_test_split(X,y,test_size=0.3,random_state=15)
print("training set:",X_train.shape,'-',y_train.shape[0],'samples')
print("testing set:",X_test.shape,'-',y_test.shape[0], 'samples')
     training set: (12774, 8) - 12774 samples
     testing set: (5475, 8) - 5475 samples
# build and fit the model
from sklearn.preprocessing import StandardScaler
# Create an instance of StandardScaler
scaler = StandardScaler()
# Fit the scaler to your training data
scaler.fit(X_train)
# Transform the training and test data using the scaler
X_train_scaled = scaler.transform(X_train)
X_test_scaled = scaler.transform(X_test)
# Build and fit the linear regression model
model = LinearRegression()
model.fit(X_train_scaled, y_train)
      ▼ LinearRegression
     LinearRegression()
# Evaluate the Results:
# Prediction and Calculate the accuracy for the testing dataset
test_pre = model.predict(X_test)
test_score = r2_score(y_test,test_pre)
print("The accuracy of testing dataset",test_score*100)
     The accuracy of testing dataset -1.4505457449012067e+23
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:432: UserWarning: X has feature names, but LinearRegression wa
       warnings.warn(
# Prediction and Calculate the accuracy for the training dataset
train_pre = model.predict(X_train)
train_score = r2_score(y_train,train_pre)
```

print("The accuracy of training dataset",train_score*100)

The accuracy of training dataset -1.1461742718581598e+23/usr/local/lib/python3.10/dist-packages/sklearn/base.py:432: UserWarning: X has feature names, but LinearRegression wa warnings.warn(

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