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

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

1: import pandas as pd

1: ava=pd.read_csv("/content/drive/MyDrive/STUDY2/DATA ANALYSIS LAB/LABCYCLE/DATASETS/avocado_10thQuestion.csv")

1: ava

1: 
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	Unnamed: 0	Date	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags	Large Bags	XLarge Bags	type	year	region
0	0	12/27/2015	1.33	64236.62	1036.74	54454.85	48.16	8696.87	8603.62	93.25	0.0	conventional	2015	Albany
1	1	12/20/2015	1.35	54876.98	674.28	44638.81	58.33	9505.56	9408.07	97.49	0.0	conventional	2015	Albany
2	2	12/13/2015	0.93	118220.22	794.70	109149.67	130.50	8145.35	8042.21	103.14	0.0	conventional	2015	Albany
3	3	12/6/2015	1.08	78992.15	1132.00	71976.41	72.58	5811.16	5677.40	133.76	0.0	conventional	2015	Albany
4	4	11/29/2015	1.28	51039.60	941.48	43838.39	75.78	6183.95	5986.26	197.69	0.0	conventional	2015	Albany
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
18244	7	2/4/2018	1.63	17074.83	2046.96	1529.20	0.00	13498.67	13066.82	431.85	0.0	organic	2018	WestTexNewMexico
18245	8	1/28/2018	1.71	13888.04	1191.70	3431.50	0.00	9264.84	8940.04	324.80	0.0	organic	2018	WestTexNewMexico
18246	9	1/21/2018	1.87	13766.76	1191.92	2452.79	727.94	9394.11	9351.80	42.31	0.0	organic	2018	WestTexNewMexico
18247	10	1/14/2018	1.93	16205.22	1527.63	2981.04	727.01	10969.54	10919.54	50.00	0.0	organic	2018	WestTexNewMexico
18248	11	1/7/2018	1.62	17489.58	2894.77	2356.13	224.53	12014.15	11988.14	26.01	0.0	organic	2018	WestTexNewMexico

18249 rows x 14 columns

```
1: #a) How to identify the unique values in the region column.
   pd.Series(ava["region"]).unique()

1: 
```

- Albany
- Atlanta
- BaltimoreWashington
- Boise
- Boston
- BuffaloRochester
- California
- Charlotte
- Chicago
- CincinnatiDayton
- Columbus
- DallasFtWorth
- Denver
- Detroit
- GrandRapids
- GreatLakes
- HarrisburgScranton
- HartfordSpringfield
- Houston
- Indianapolis
- Jacksonville
- LasVegas
- LosAngeles
- Louisville
- MiamiFtLauderdale
- Midsouth
- Nashville
- NewOrleansMobile
- NewYork
- Northeast
- NorthernNewEngland
- Orlando
- Philadelphia
- PhoenixTucson
- Pittsburgh
- Plains
- Portland
- RaleighGreensboro
- RichmondNorfolk
- Roanoke
- Sacramento
- SanDiego
- SanFrancisco
- Seattle
- SouthCarolina
- SouthCentral
- Southeast
- Spokane
- StLouis
- Syracuse
- Tampa

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51         TotalUS
52         West
53         WestTexNewMexico
dtype: object

1: #b) What is the maximum price for an avocado in the dataset.
ava["AveragePrice"].max()

1: 3.25

1: #c) Identify the type distribution and take a single avocado in the dataset and find out the median price ,mean, and standard deviation.
print(ava["type"].value_counts())
single=ava[ava["type"]=="conventional"]
print(single["AveragePrice"].mean())
print(single["AveragePrice"].median())
print(single["AveragePrice"].std())

1: conventional    9126
organic          9123
Name: type, dtype: int64
1.1580396668858208
1.13
0.26304060411401714

1: #d) Find the highest, lowest price for conventional avocado's in year with location.
conv=ava[ava["type"]=="conventional"]
p=conv.groupby(['year', 'region'])["AveragePrice"]
print(p.max())
print(p.min())

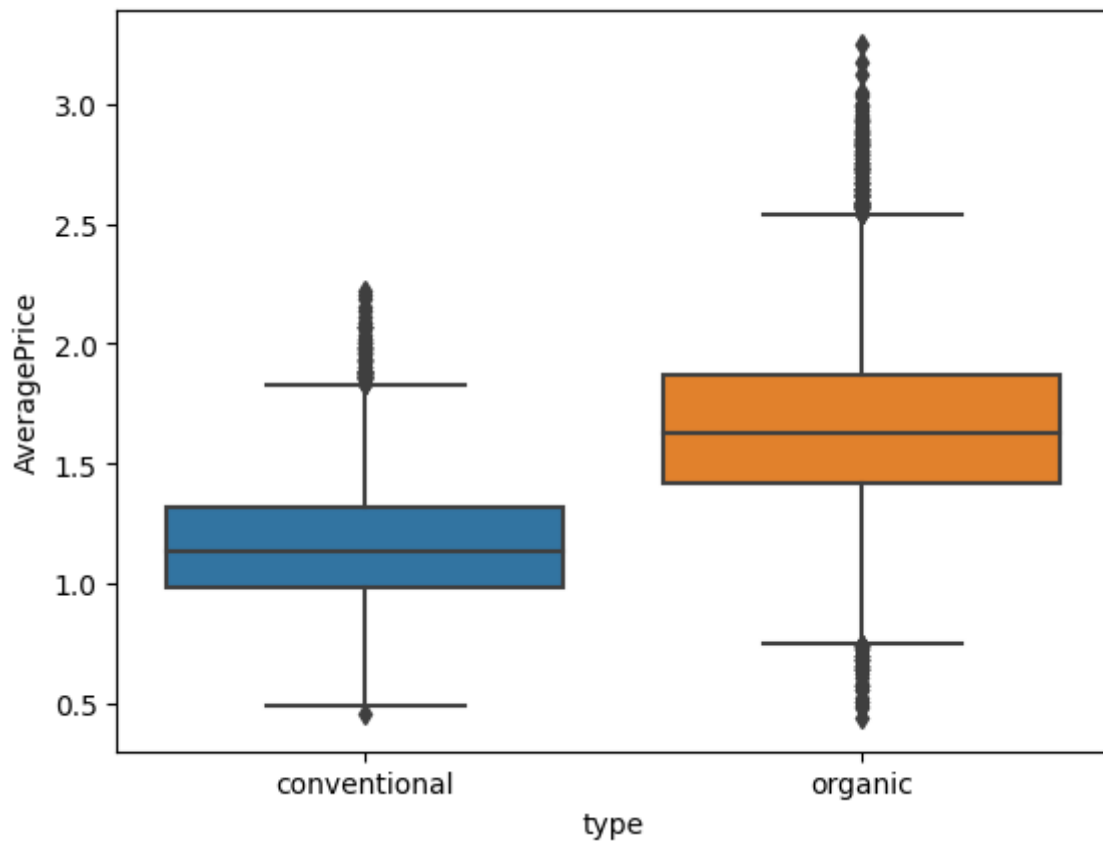
1: year region
2015 Albany 1.45
Atlanta 1.15
BaltimoreWashington 1.29
Boise 1.19
Boston 1.28
...
2018 Syracuse 1.45
Tampa 1.41
TotalUS 1.20
West 1.08
WestTexNewMexico 0.94
Name: AveragePrice, Length: 216, dtype: float64
year region
2015 Albany 0.93
Atlanta 0.93
BaltimoreWashington 0.99
Boise 0.71
Boston 0.94
...
2018 Syracuse 1.10
Tampa 0.98
TotalUS 0.87
West 0.83
WestTexNewMexico 0.75
Name: AveragePrice, Length: 216, dtype: float64

1: #e) Draw the plots of the distribution of average price for different types of Avocados
import seaborn as sns
sns.boxplot(x='type', y='AveragePrice', data=ava)

1:

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1:



1: #f) Find the correlation matrix to measure the strength of the correlation between variables.  
ava.select\_dtypes(include=['float64', 'int64']).corr()

	Unnamed: 0	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags	Large Bags	XLarge Bags	year
Unnamed: 0	1.000000	-0.133008	0.014035	0.017628	0.019829	0.041752	-0.002219	0.000347	-0.009196	-0.011546	-0.171667
AveragePrice	-0.133008	1.000000	-0.192752	-0.208317	-0.172928	-0.179446	-0.177088	-0.174730	-0.172940	-0.117592	0.093197
Total Volume	0.014035	-0.192752	1.000000	0.977863	0.974181	0.872202	0.963047	0.967238	0.880640	0.747157	0.017193
4046	0.017628	-0.208317	0.977863	1.000000	0.926110	0.833389	0.920057	0.925280	0.838645	0.699377	0.003353
4225	0.019829	-0.172928	0.974181	0.926110	1.000000	0.887855	0.905787	0.916031	0.810015	0.688809	-0.009559
4770	0.041752	-0.179446	0.872202	0.833389	0.887855	1.000000	0.792314	0.802733	0.698471	0.679861	-0.036531
Total Bags	-0.002219	-0.177088	0.963047	0.920057	0.905787	0.792314	1.000000	0.994335	0.943009	0.804233	0.071552
Small Bags	0.000347	-0.174730	0.967238	0.925280	0.916031	0.802733	0.994335	1.000000	0.902589	0.806845	0.063915
Large Bags	-0.009196	-0.172940	0.880640	0.838645	0.810015	0.698471	0.943009	0.902589	1.000000	0.710858	0.087891
XLarge Bags	-0.011546	-0.117592	0.747157	0.699377	0.688809	0.679861	0.804233	0.806845	0.710858	1.000000	0.081033
year	-0.171667	0.093197	0.017193	0.003353	-0.009559	-0.036531	0.071552	0.063915	0.087891	0.081033	1.000000

1: #g) Find out the volume of avocado sales has increased in the last 5 years.  
current\_year = ava['year'].max()  
five\_years\_ago = current\_year - 5  
  
last\_5\_years\_volume = ava[ava['year'] >= five\_years\_ago]['Total Volume'].sum()  
total\_volume = ava['Total Volume'].sum()  
last\_5\_years\_volume - total\_volume

1: 0.0