

# updated assignment

March 8, 2022

Project Title:Vehicle Sales Analysis

Authors: Mostafa Abdelazim, Zeina Kishk, and Hania Raslan

Abstract: This dataset describes the vehicle market in prospective of sales depending on several factors: pricing,timing, location(city,and country), and describes how the product reached the consumer in addition to the process of shipping. We obtained the dataset from kaggle.

```
[3]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.api as sm
from statsmodels.graphics.mosaicplot import mosaic
```

```
[2]: data = pd.read_csv("dataset.csv", encoding= 'unicode_escape')
print(data.head())
print(data.tail())
```

	ORDERNUMBER	QUANTITYORDERED	PRICEEACH	ORDERLINENUMBER	SALES	STATUS	\
0	10107	30	95.70	2	2871.00	Shipped	
1	10121	34	81.35	5	2765.90	Shipped	
2	10134	41	94.74	2	3884.34	Shipped	
3	10145	45	83.26	6	3746.70	Shipped	
4	10159	49	100.00	14	5205.27	Shipped	

	QTR_ID	MONTH_ID	YEAR_ID	PRODUCTLINE	...	PHONE	\
0	1	2	2003	Motorcycles	...	2125557818	
1	2	5	2003	Motorcycles	...	26.47.1555	
2	3	7	2003	Motorcycles	...	+33 1 46 62 7555	
3	3	8	2003	Motorcycles	...	6265557265	
4	4	10	2003	Motorcycles	...	6505551386	

	ADDRESSLINE1	CITY	STATE	POSTALCODE	COUNTRY	\
0	897 Long Airport Avenue	NYC	NY	10022	USA	
1	59 rue de l'Abbaye	Reims	NaN	51100	France	
2	27 rue du Colonel Pierre Avia	Paris	NaN	75508	France	
3	78934 Hillside Dr.	Pasadena	CA	90003	USA	
4	7734 Strong St.	San Francisco	CA	NaN	USA	

	TERRITORY	CONTACTLASTNAME	CONTACTFIRSTNAME	DEALSIZE
0	NaN	Yu	Kwai	Small
1	EMEA	Henriot	Paul	Small
2	EMEA	Da Cunha	Daniel	Medium
3	NaN	Young	Julie	Medium
4	NaN	Brown	Julie	Medium

[5 rows x 23 columns]

	ORDERNUMBER	QUANTITYORDERED	PRICEEACH	ORDERLINENUMBER	SALES	\
2818	10350	20	100.00	15	2244.40	
2819	10373	29	100.00	1	3978.51	
2820	10386	43	100.00	4	5417.57	
2821	10397	34	62.24	1	2116.16	
2822	10414	47	65.52	9	3079.44	

	STATUS	QTR_ID	MONTH_ID	YEAR_ID	PRODUCTLINE	...	PHONE	\
2818	Shipped	4	12	2004	Ships	...	(91) 555 94 44	
2819	Shipped	1	1	2005	Ships	...	981-443655	
2820	Resolved	1	3	2005	Ships	...	(91) 555 94 44	
2821	Shipped	1	3	2005	Ships	...	61.77.6555	
2822	On Hold	2	5	2005	Ships	...	6175559555	

	ADDRESSLINE1	CITY	STATE	POSTALCODE	COUNTRY	TERRITORY	\
2818	C/ Moralarzal, 86	Madrid	NaN	28034	Spain	EMEA	
2819	Torikatu 38	Oulu	NaN	90110	Finland	EMEA	
2820	C/ Moralarzal, 86	Madrid	NaN	28034	Spain	EMEA	
2821	1 rue Alsace-Lorraine	Toulouse	NaN	31000	France	EMEA	
2822	8616 Spinnaker Dr.	Boston	MA	51003	USA	NaN	

	CONTACTLASTNAME	CONTACTFIRSTNAME	DEALSIZE
2818	Freyre	Diego	Small
2819	Koskitalo	Pirkko	Medium
2820	Freyre	Diego	Medium
2821	Roulet	Annette	Small
2822	Yoshido	Juri	Medium

[5 rows x 23 columns]

```
[3]: ##Mean
print(data['PRICEEACH'].mean())
print(data['SALES'].mean())
print(data['QUANTITYORDERED'].mean())
print(data['ORDERLINENUMBER'].mean())
print(data['MSRP'].mean())
```

83.65854410201929

3553.88907190932

35.09280906836698

6.466170740347148  
100.71555083244775

```
[4]: #Median
print(data['PRICEEACH'].median())
print(data['SALES'].median())
print(data['QUANTITYORDERED'].median())
print(data['ORDERLINENUMBER'].median())
print(data['MSRP'].median())
```

95.7  
3184.8  
35.0  
6.0  
99.0

```
[5]: #Variance and Std
print(data['ORDERLINENUMBER'].var())
print(data['ORDERLINENUMBER'].std())
print(data['MSRP'].var())
print(data['MSRP'].std())
print(data['PRICEEACH'].var())
print(data['PRICEEACH'].std())
print(data['QUANTITYORDERED'].var())
print(data['QUANTITYORDERED'].std())
print(data['SALES'].var())
print(data['SALES'].std())
```

17.85773185886007  
4.225840964690942  
1615.0682449746778  
40.187911677203104  
407.0014334217844  
20.174276527840703  
94.89570659960627  
9.74144273706961  
3392467.0677432865  
1841.8651057401805

```
[6]: #Describe (Quantiles)
data.describe()
```

```
[6]:
```

	ORDERNUMBER	QUANTITYORDERED	PRICEEACH	ORDERLINENUMBER	\
count	2823.000000	2823.000000	2823.000000	2823.000000	
mean	10258.725115	35.092809	83.658544	6.466171	
std	92.085478	9.741443	20.174277	4.225841	
min	10100.000000	6.000000	26.880000	1.000000	
25%	10180.000000	27.000000	68.860000	3.000000	
50%	10262.000000	35.000000	95.700000	6.000000	

75%	10333.500000	43.000000	100.000000	9.000000
max	10425.000000	97.000000	100.000000	18.000000

	SALES	QTR_ID	MONTH_ID	YEAR_ID	MSRP
count	2823.000000	2823.000000	2823.000000	2823.000000	2823.000000
mean	3553.889072	2.717676	7.092455	2003.81509	100.715551
std	1841.865106	1.203878	3.656633	0.69967	40.187912
min	482.130000	1.000000	1.000000	2003.00000	33.000000
25%	2203.430000	2.000000	4.000000	2003.00000	68.000000
50%	3184.800000	3.000000	8.000000	2004.00000	99.000000
75%	4508.000000	4.000000	11.000000	2004.00000	124.000000
max	14082.800000	4.000000	12.000000	2005.00000	214.000000

```
[7]: #Discrete Values frequency
print(data['PRODUCTLINE'].value_counts(normalize=True))
print(data['MONTH_ID'].value_counts(normalize=True))
print(data['STATUS'].value_counts(normalize=True))
print(data['CITY'].value_counts(normalize=True))
print(data['COUNTRY'].value_counts(normalize=True))
```

```
Classic Cars      0.342543
Vintage Cars     0.215019
Motorcycles      0.117251
Planes           0.108395
Trucks and Buses 0.106624
Ships            0.082891
Trains           0.027276
Name: PRODUCTLINE, dtype: float64
11    0.211477
10    0.112292
5     0.089267
1     0.081119
2     0.079348
3     0.075097
8     0.067659
12    0.063762
4     0.063053
9     0.060574
7     0.049947
6     0.046405
Name: MONTH_ID, dtype: float64
Shipped      0.927028
Cancelled    0.021254
Resolved     0.016649
On Hold      0.015586
In Process   0.014524
Disputed     0.004959
Name: STATUS, dtype: float64
```

Madrid	0.107687
San Rafael	0.063762
NYC	0.053843
Singapore	0.027984
Paris	0.024796
...	
Graz	0.005313
Los Angeles	0.004959
Munich	0.004959
Burbank	0.004605
Charleroi	0.002834

Name: CITY, Length: 73, dtype: float64

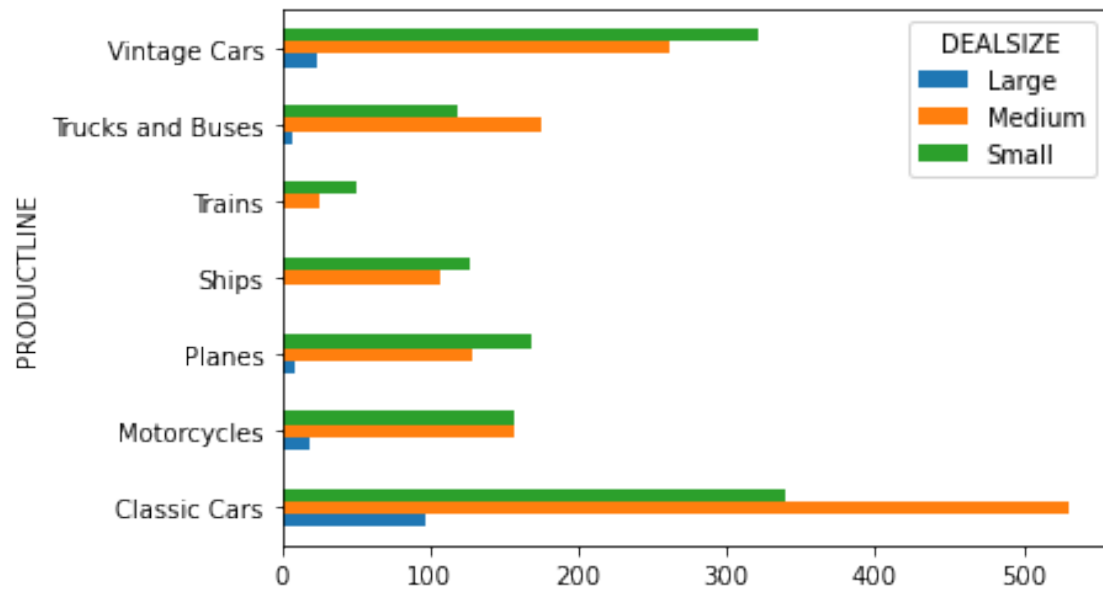
USA	0.355650
Spain	0.121148
France	0.111229
Australia	0.065533
UK	0.051010
Italy	0.040028
Finland	0.032589
Norway	0.030110
Singapore	0.027984
Canada	0.024796
Denmark	0.022317
Germany	0.021962
Sweden	0.020191
Austria	0.019483
Japan	0.018420
Belgium	0.011690
Switzerland	0.010981
Philippines	0.009210
Ireland	0.005668

Name: COUNTRY, dtype: float64

```
[8]: #Grouping by a variable
grouped = data.groupby('PRODUCTLINE')
```

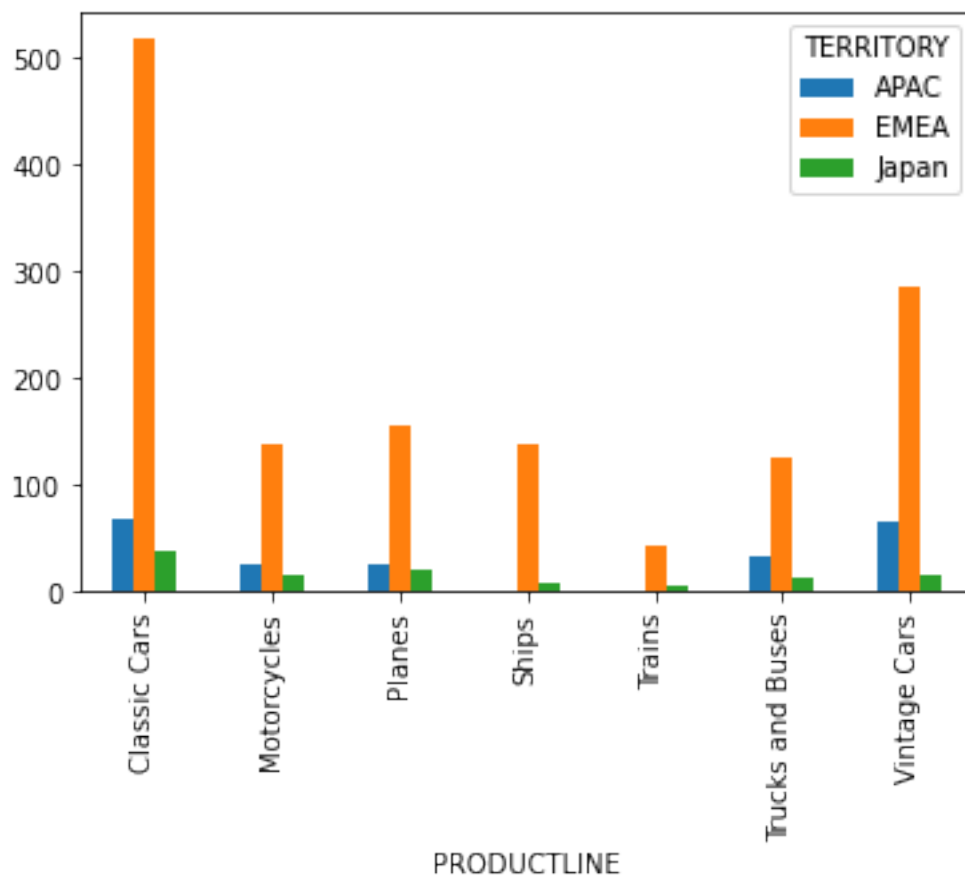
```
[9]: #Barplot 1
grouped['DEALSIZE'].value_counts().unstack().plot(kind="barh")
```

```
[9]: <AxesSubplot:ylabel='PRODUCTLINE'>
```



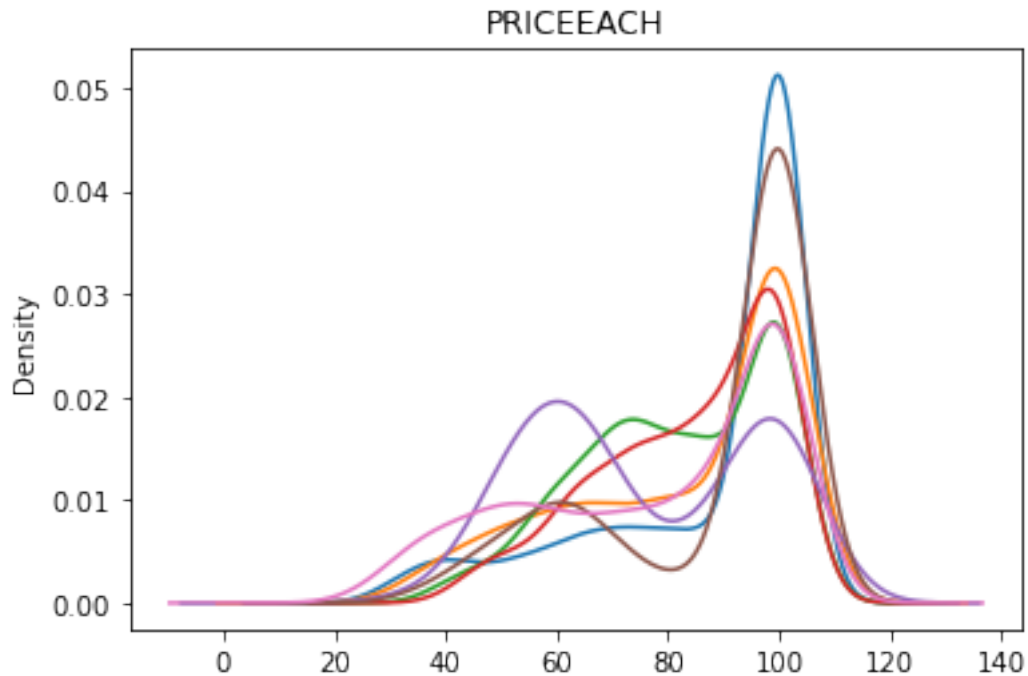
```
[10]: #Barplot 2
grouped['TERRITORY'].value_counts().unstack().plot(kind="bar")
```

```
[10]: <AxesSubplot:xlabel='PRODUCTLINE'>
```



```
[11]: #Density Plot 1
grouped['PRICEEACH'].plot(kind='density', title='PRICEEACH')
```

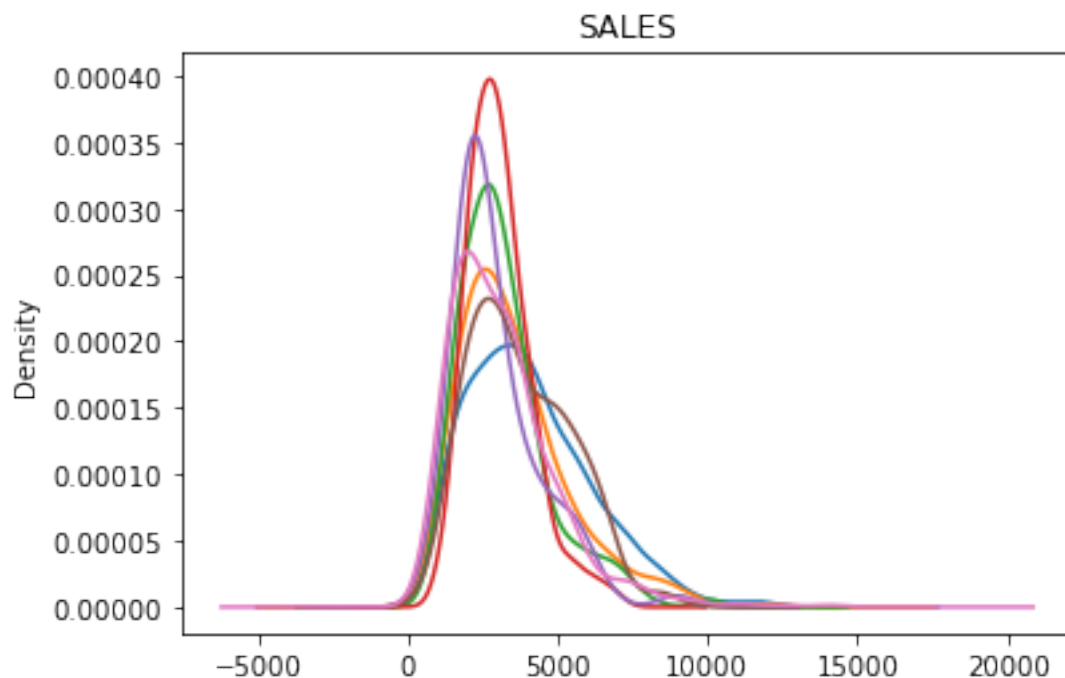
```
[11]: PRODUCTLINE
Classic Cars      AxesSubplot(0.125,0.125;0.775x0.755)
Motorcycles       AxesSubplot(0.125,0.125;0.775x0.755)
Planes            AxesSubplot(0.125,0.125;0.775x0.755)
Ships             AxesSubplot(0.125,0.125;0.775x0.755)
Trains            AxesSubplot(0.125,0.125;0.775x0.755)
Trucks and Buses  AxesSubplot(0.125,0.125;0.775x0.755)
Vintage Cars      AxesSubplot(0.125,0.125;0.775x0.755)
Name: PRICEEACH, dtype: object
```



```
[12]: #Density Plot 2
grouped['SALES'].plot(kind='density', title='SALES')
```

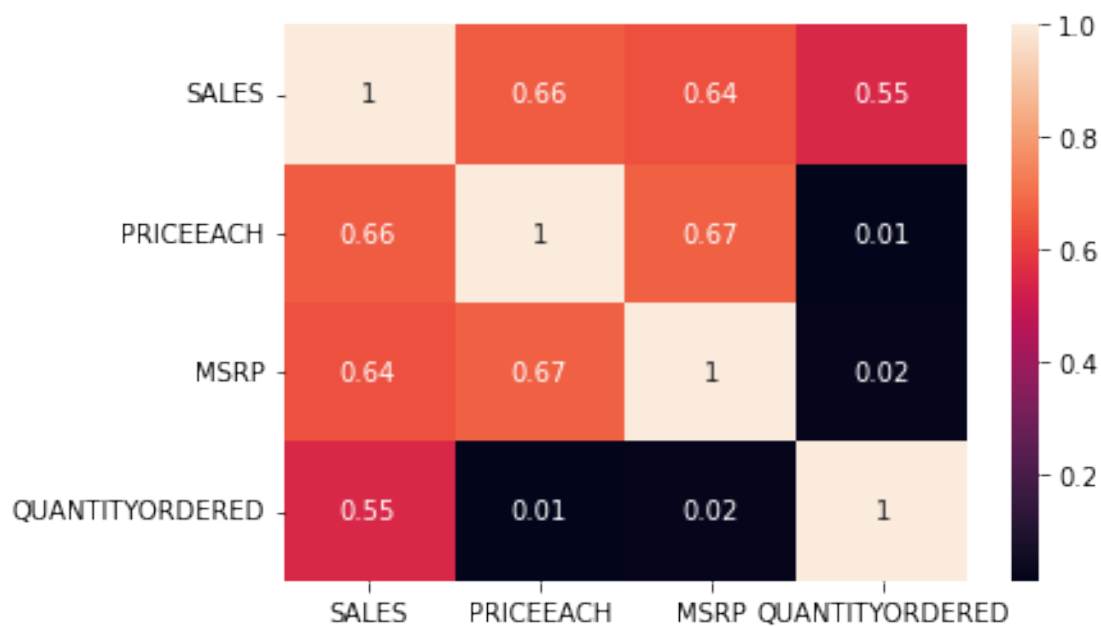
```
[12]: PRODUCTLINE
Classic Cars      AxesSubplot(0.125,0.125;0.775x0.755)
Motorcycles      AxesSubplot(0.125,0.125;0.775x0.755)
Planes           AxesSubplot(0.125,0.125;0.775x0.755)
Ships            AxesSubplot(0.125,0.125;0.775x0.755)
Trains           AxesSubplot(0.125,0.125;0.775x0.755)
Trucks and Buses AxesSubplot(0.125,0.125;0.775x0.755)
Vintage Cars     AxesSubplot(0.125,0.125;0.775x0.755)
Name: SALES, dtype: object
```





```
[13]: #Heat Map 1
cormat=data[['SALES','PRICEEACH','MSRP','QUANTITYORDERED']].corr().round(2)
sns.heatmap(cormat, annot=True)
```

[13]: <AxesSubplot:>



```
[14]: # CORRELATION
print(data.corr())
# CORRELATION BETWEEN SALES AND PRICE EACH
print(data[['SALES', 'PRICEEACH']].corr(method='pearson'))
# CORRELATION BETWEEN SALES AND MSRP
print(data[['MSRP', 'SALES']].corr(method='pearson'))
# CORRELATION BETWEEN MSRP AND PRICE EACH
print(data[['MSRP', 'PRICEEACH']].corr(method='pearson'))
# CORRELATION BETWEEN QUANTITYORDERED AND PRICE EACH
print(data[['QUANTITYORDERED', 'PRICEEACH']].corr(method='pearson'))
# CORRELATION BETWEEN SALES AND QUANTITYORDERED
print(data[['QUANTITYORDERED', 'SALES']].corr(method='pearson'))
```

	ORDERNUMBER	QUANTITYORDERED	PRICEEACH	ORDERLINENUMBER	\
ORDERNUMBER	1.000000	0.065543	-0.002935	-0.055550	
QUANTITYORDERED	0.065543	1.000000	0.005564	-0.018397	
PRICEEACH	-0.002935	0.005564	1.000000	-0.020965	
ORDERLINENUMBER	-0.055550	-0.018397	-0.020965	1.000000	
SALES	0.039919	0.551426	0.657841	-0.058400	
QTR_ID	-0.051383	-0.035323	0.008712	0.040716	
MONTH_ID	-0.039723	-0.039048	0.005152	0.034016	
YEAR_ID	0.904596	0.069535	-0.005938	-0.057367	
MSRP	-0.010280	0.017881	0.670625	-0.021067	

	SALES	QTR_ID	MONTH_ID	YEAR_ID	MSRP
ORDERNUMBER	0.039919	-0.051383	-0.039723	0.904596	-0.010280
QUANTITYORDERED	0.551426	-0.035323	-0.039048	0.069535	0.017881
PRICEEACH	0.657841	0.008712	0.005152	-0.005938	0.670625
ORDERLINENUMBER	-0.058400	0.040716	0.034016	-0.057367	-0.021067
SALES	1.000000	-0.006796	-0.009605	0.035647	0.635239
QTR_ID	-0.006796	1.000000	0.979300	-0.433052	0.010234
MONTH_ID	-0.009605	0.979300	1.000000	-0.430163	0.008170
YEAR_ID	0.035647	-0.433052	-0.430163	1.000000	-0.014310
MSRP	0.635239	0.010234	0.008170	-0.014310	1.000000

	SALES	PRICEEACH
SALES	1.000000	0.657841
PRICEEACH	0.657841	1.000000

	MSRP	SALES
MSRP	1.000000	0.635239
SALES	0.635239	1.000000

	MSRP	PRICEEACH
MSRP	1.000000	0.670625
PRICEEACH	0.670625	1.000000

	QUANTITYORDERED	PRICEEACH
QUANTITYORDERED	1.000000	0.005564

PRICEEACH	0.005564	1.000000
	QUANTITYORDERED	SALES
QUANTITYORDERED	1.000000	0.551426
SALES	0.551426	1.000000

For the correlation between SALES and PRICEEACH, we can see that they have a positive correlation. For the correlation between SALES and MSRP, we can see that they also have a positive correlation which is very similar to the one between SALES and PRICEEACH which may be considered moderate (not strong but not weak). For the correlation between MSRP and PRICEEACH, it is also a positive moderate correlation. For the correlation between QUANTITYORDERED and PRICEEACH, we can see that they have a very weak positive correlation. Lastly, for the correlation between QUANTITYORDERED and SALES, the correlation is also moderate and positive.

```
[15]: #Cross Tables (contingency table)
tab = pd.crosstab(data['SALES'], data['PRICEEACH'])
print(tab)
tab1 = pd.crosstab(data['SALES'], data['MSRP'])
print(tab1)
tab2 = pd.crosstab(data['PRODUCTLINE'], data['SALES'])
print(tab2)
tab3 = pd.crosstab(data['SALES'], data['QUANTITYORDERED'])
print(tab3)
tab4 = pd.crosstab(data['MSRP'], data['QUANTITYORDERED'])
print(tab4)
```

PRICEEACH	26.88	27.22	28.29	28.88	29.21	29.54	29.70	29.87	\
SALES									
482.13	0	0	0	0	0	0	0	0	
541.14	0	0	0	0	0	0	0	0	
553.95	0	0	0	0	0	0	0	0	
577.60	0	0	0	1	0	0	0	0	
640.05	0	0	0	0	0	0	0	0	
...	...	...	...	...	...	...	...	...	
11886.60	0	0	0	0	0	0	0	0	
11887.80	0	0	0	0	0	0	0	0	
12001.00	0	0	0	0	0	0	0	0	
12536.50	0	0	0	0	0	0	0	0	
14082.80	0	0	0	0	0	0	0	0	

PRICEEACH	30.06	30.20	...	99.55	99.57	99.58	99.66	99.67	\
SALES			...						
482.13	0	0	...	0	0	0	0	0	
541.14	0	0	...	0	0	0	0	0	
553.95	0	0	...	0	0	0	0	0	
577.60	0	0	...	0	0	0	0	0	
640.05	0	0	...	0	0	0	0	0	
...	...	...	...	...	...	...	...	...	
11886.60	0	0	...	0	0	0	0	0	
11887.80	0	0	...	0	0	0	0	0	

12001.00	0	0	...	0	0	0	0	0
12536.50	0	0	...	0	0	0	0	0
14082.80	0	0	...	0	0	0	0	0

PRICEEACH	99.69	99.72	99.82	99.91	100.00
-----------	-------	-------	-------	-------	--------

SALES

482.13	0	0	0	0	0
541.14	0	0	0	0	0
553.95	0	0	0	0	0
577.60	0	0	0	0	0
640.05	0	0	0	0	0

...	...	...	...	...	...
11886.60	0	0	0	0	1
11887.80	0	0	0	0	1
12001.00	0	0	0	0	1
12536.50	0	0	0	0	1
14082.80	0	0	0	0	1

[2763 rows x 1016 columns]

MSRP	33	35	37	40	41	43	44	49	50	53	...	157	163	\
------	----	----	----	----	----	----	----	----	----	----	-----	-----	-----	---

SALES

482.13	0	0	0	0	0	0	0	0	0	0	...	0	0
541.14	0	0	0	0	0	0	0	0	0	0	...	0	0
553.95	0	0	0	0	1	0	0	0	0	0	...	0	0
577.60	1	0	0	0	0	0	0	0	0	0	...	0	0
640.05	0	0	1	0	0	0	0	0	0	0	...	0	0

...	...	...	...	...	...	...	...	...	...	...	...	...	...
11886.60	0	0	0	0	0	0	0	0	0	0	...	0	0
11887.80	0	0	0	0	0	0	0	0	0	0	...	0	0
12001.00	0	0	0	0	0	0	0	0	0	0	...	0	0
12536.50	0	0	0	0	0	0	0	0	0	0	...	0	0
14082.80	0	0	0	0	0	0	0	0	0	0	...	0	0

MSRP	168	169	170	173	193	194	207	214
------	-----	-----	-----	-----	-----	-----	-----	-----

SALES

482.13	0	0	0	0	0	0	0	0
541.14	0	0	0	0	0	0	0	0
553.95	0	0	0	0	0	0	0	0
577.60	0	0	0	0	0	0	0	0
640.05	0	0	0	0	0	0	0	0

...	...	...	...	...	...	...	...
11886.60	0	0	0	0	1	0	0
11887.80	0	1	0	0	0	0	0
12001.00	0	0	0	0	0	0	1
12536.50	0	0	0	0	0	0	0
14082.80	0	0	1	0	0	0	0

[2763 rows x 80 columns]

SALES	482.13	541.14	553.95	577.60	640.05	651.80	\
PRODUCTLINE							
Classic Cars	0	0	0	0	1	0	
Motorcycles	0	0	0	0	0	1	
Planes	0	0	0	0	0	0	
Ships	0	0	0	0	0	0	
Trains	0	0	0	0	0	0	
Trucks and Buses	1	0	0	0	0	0	
Vintage Cars	0	1	1	1	0	0	

SALES	652.35	683.80	694.60	703.60	...	10993.50	\
PRODUCTLINE					...		
Classic Cars	0	0	0	0	...	1	
Motorcycles	0	0	0	0	...	0	
Planes	0	0	0	0	...	0	
Ships	0	0	0	0	...	0	
Trains	0	0	0	0	...	0	
Trucks and Buses	0	0	0	0	...	0	
Vintage Cars	1	1	1	1	...	0	

SALES	11279.20	11336.70	11623.70	11739.70	11886.60	11887.80	\
PRODUCTLINE							
Classic Cars	1	0	1	1	0	1	
Motorcycles	0	0	0	0	1	0	
Planes	0	0	0	0	0	0	
Ships	0	0	0	0	0	0	
Trains	0	0	0	0	0	0	
Trucks and Buses	0	0	0	0	0	0	
Vintage Cars	0	1	0	0	0	0	

SALES	12001.00	12536.50	14082.80
PRODUCTLINE			
Classic Cars	1	0	0
Motorcycles	0	0	0
Planes	0	0	0
Ships	0	0	0
Trains	0	0	0
Trucks and Buses	0	0	0
Vintage Cars	0	1	1

[7 rows x 2763 columns]

QUANTITYORDERED	6	10	11	12	13	15	16	18	19	20	...	61	62	64	65	\
SALES											...					
482.13	0	0	1	0	0	0	0	0	0	0	...	0	0	0	0	
541.14	1	0	0	0	0	0	0	0	0	0	...	0	0	0	0	
553.95	0	0	0	0	0	1	0	0	0	0	...	0	0	0	0	
577.60	0	0	0	0	0	0	0	0	0	1	...	0	0	0	0	
640.05	0	0	0	0	0	1	0	0	0	0	...	0	0	0	0	

...	..	..	..	..	..	..	..	..	..	..	...	..	..	..	..
11886.60	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0
11887.80	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0
12001.00	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0
12536.50	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0
14082.80	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0

QUANTITYORDERED	66	70	76	77	85	97
SALES						
482.13	0	0	0	0	0	0
541.14	0	0	0	0	0	0
553.95	0	0	0	0	0	0
577.60	0	0	0	0	0	0
640.05	0	0	0	0	0	0

...	..	..	..	..	..	..
11886.60	1	0	0	0	0	0
11887.80	0	0	0	0	0	0
12001.00	0	0	0	0	0	0
12536.50	0	0	0	0	0	0
14082.80	0	0	1	0	0	0

[2763 rows x 58 columns]

QUANTITYORDERED	6	10	11	12	13	15	16	18	19	20	...	61	62	64	65	\
MSRP											...					
33	0	0	0	0	0	0	0	0	0	5	...	1	0	0	0	
35	0	0	0	0	0	0	0	0	0	1	...	0	0	0	0	
37	0	0	0	0	0	1	0	0	0	2	...	0	0	0	0	
40	0	0	0	0	0	0	0	0	0	1	...	0	0	0	0	
41	0	0	0	0	0	2	0	0	0	0	...	0	0	0	0	
...	..	..	..	..	..	..	..	..	..	..	...	..	..	..	..	..
173	0	0	0	0	0	0	0	0	0	2	...	0	0	0	0	
193	0	0	0	0	0	0	0	0	0	2	...	0	0	0	0	
194	0	0	0	1	0	0	0	0	0	1	...	0	0	0	0	
207	0	0	0	0	0	0	0	0	0	1	...	0	0	0	0	
214	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	

QUANTITYORDERED	66	70	76	77	85	97
MSRP						
33	0	0	0	0	0	0
35	0	0	0	0	0	0
37	0	0	0	0	0	0
40	0	0	0	0	0	0
41	0	0	0	0	0	0
...	..	..	..	..	..	..
173	0	0	0	0	0	0
193	1	0	0	0	0	0
194	0	0	0	0	0	0
207	0	0	0	0	0	0

```
214          0  0  0  0  0  0
```

```
[80 rows x 58 columns]
```

```
[16]: #Mosaic Graph 1
mosaic(data, ['DEALSIZE', 'TERRITORY'], title=' Deal size x Territory ')
plt.show()
```

