

```
In [3]: import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
%matplotlib inline
import plotly.express as px
import seaborn as sns
```

```
In [ ]:
```

```
In [4]: df=pd.read_excel("Copy of Budget data1.xlsx")
```

```
In [5]: df
```

Out[5]:

	Category	Subcategory	ProductName	ProductKey	Jan, 2016	Feb, 2016	Mar, 2016	Apr, 2016	May, 2016	Jun, 2016	Jul, 2016	Aug, 2016	Sep, 2016	Oct, 2016
0	Accessories	Bike Racks	Hitch Rack - 4-Bike	483.0	1131	2635	4134	2179	2637	3279	2218	3287	3885	2484
1	Accessories	Bike Stands	All-Purpose Bike Stand	486.0	666	3695	2868	4862	3439	4612	2774	3003	2401	4413
2	Accessories	Bottles and Cages	Water Bottle - 30 oz.	477.0	1892	4727	3656	4449	4051	6257	4871	5231	5461	5529
3	Accessories	Cleaners	Bike Wash - Dissolver	484.0	160	713	555	656	369	582	777	777	239	496
4	Accessories	Fenders	Fender Set - Mountain	485.0	970	3014	2809	4259	3638	3721	4190	3618	3975	3892
5	Accessories	Helmets	Sport-100 Helmet, Red	212.0	5317	16221	16752	16552	17204	25354	17584	20409	18268	20567
6	Accessories	Hydration Packs	Hydration Pack - 70 oz.	487.0	809	2684	2917	3425	2716	3260	3773	3523	4252	3111
7	Accessories	Tires and Tubes	Patch Kit/8 Patches	480.0	3554	18758	20905	18046	21680	22456	23995	22922	20950	21905
8	SubTotal Accessories	NaN	NaN	NaN	14499	52447	54596	54428	55734	69521	60182	62770	59431	62397
9	Bikes	Mountain Bikes	Mountain-100 Silver, 38	344.0	370105	326786	384811	439822	458523	619456	524348	647048	557368	615032
10	Bikes	Road Bikes	Road-150 Red, 62	310.0	346295	289524	355097	346783	399691	546092	441037	432400	468572	483913
11	Bikes	Touring Bikes	Touring-2000 Blue, 60	560.0	133631	165941	178287	265901	286630	445270	299106	407069	391580	481316
12	SubTotal Bikes	NaN	NaN	NaN	850031	782251	918195	1052506	1144844	1610818	1264491	1486517	1417520	1580261
13	Clothing	Caps	AWC Logo Cap	223.0	479	1695	1462	1079	1729	2180	1588	2065	2013	2138
14	Clothing	Gloves	Half-Finger Gloves, S	462.0	598	2474	2957	2705	2819	2966	2975	3264	2424	3181

	Category	Subcategory	ProductName	ProductKey	Jan, 2016	Feb, 2016	Mar, 2016	Apr, 2016	May, 2016	Jun, 2016	Jul, 2016	Aug, 2016	Sep, 2016	Oct, 2016	
15	Clothing	Jerseys	Long-Sleeve Logo Jersey, S	226.0	4087	11508	12872	11809	12789	18153	16846	13497	15988	15920	1
16	Clothing	Shorts	Men's Sports Shorts, S	445.0	421	5723	7301	6335	5288	6829	4617	5384	6277	6337	
17	Clothing	Socks	Mountain Bike Socks, M	218.0	24	244	432	547	385	372	839	487	425	335	
18	Clothing	Vests	Classic Vest, S	471.0	980	2008	1980	2312	2763	2591	3379	3580	3600	4248	
19	SubTotal Clothing	NaN	NaN	NaN	6589	23652	27004	24787	25773	33091	30244	28277	30727	32159	3
20	Grand Total	NaN	NaN	NaN	871119	858350	999795	1131721	1226351	1713430	1354917	1577564	1507678	1674817	197

In [6]: `df.duplicated().sum()`

Out[6]: 0

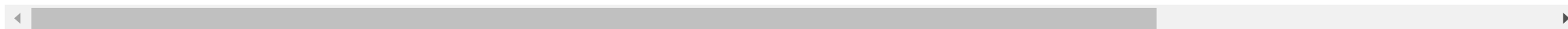
In [7]: `df1=pd.read_excel('Copy of AdventureWorks_Database.xlsx')`

In [8]: `df1`

Out[8]:

	Date	DateKey	Year	Quarter	MonthNum	Month	FiscalYear	FiscalQuarter	FiscalMonthNum	FiscalMonth	MonthYear	MonthYearLong	Month
0	2016-04-03	20160403	2016	Q2	4	Apr	FY2016	FQ4	10	Apr	Apr-16	Apr-2016	
1	2016-04-04	20160404	2016	Q2	4	Apr	FY2016	FQ4	10	Apr	Apr-16	Apr-2016	
2	2016-04-05	20160405	2016	Q2	4	Apr	FY2016	FQ4	10	Apr	Apr-16	Apr-2016	
3	2016-04-06	20160406	2016	Q2	4	Apr	FY2016	FQ4	10	Apr	Apr-16	Apr-2016	
4	2016-04-07	20160407	2016	Q2	4	Apr	FY2016	FQ4	10	Apr	Apr-16	Apr-2016	
...
1456	2014-06-18	20140618	2014	Q2	6	Jun	FY2014	FQ4	12	Jun	Jun-14	Jun-2014	
1457	2014-06-19	20140619	2014	Q2	6	Jun	FY2014	FQ4	12	Jun	Jun-14	Jun-2014	
1458	2014-06-20	20140620	2014	Q2	6	Jun	FY2014	FQ4	12	Jun	Jun-14	Jun-2014	
1459	2014-06-21	20140621	2014	Q2	6	Jun	FY2014	FQ4	12	Jun	Jun-14	Jun-2014	
1460	2014-06-22	20140622	2014	Q2	6	Jun	FY2014	FQ4	12	Jun	Jun-14	Jun-2014	

1461 rows × 16 columns

In [9]: `Customer_data=pd.read_excel('Copy of AdventureWorks Database.xlsx','Customers')`In [10]: `Customer_data`

Out[10]:

	CustomerKey	FirstName	LastName	FullName	BirthDate	MaritalStatus	Gender	YearlyIncome	TotalChildren	NumberChildrenAtHome	Educatio
0	11000	Jon	Yang	Yang, Jon	1966-04-08	M	M	90000	2	0	Bachelo
1	11001	Eugene	Huang	Huang, Eugene	1965-05-14	S	M	60000	3	3	Bachelo
2	11002	Ruben	Torres	Torres, Ruben	1965-08-12	M	M	60000	3	3	Bachelo
3	11003	Christy	Zhu	Zhu, Christy	1968-02-15	S	F	70000	0	0	Bachelo
4	11004	Elizabeth	Johnson	Johnson, Elizabeth	1968-08-08	S	F	80000	5	5	Bachelo
...
18479	29479	Tommy	Tang	Tang, Tommy	1958-07-04	M	M	30000	1	0	Graduat Degree
18480	29480	Nina	Raji	Raji, Nina	1960-11-10	S	F	30000	3	0	Graduat Degree
18481	29481	Ivan	Suri	Suri, Ivan	1960-01-05	S	M	30000	3	0	Graduat Degree
18482	29482	Clayton	Zhang	Zhang, Clayton	1959-03-05	M	M	30000	3	0	Bachelo
18483	29483	Jésus	Navarro	Navarro, Jésus	1959-12-08	M	M	30000	0	0	Bachelo

18484 rows × 17 columns



```
In [11]: Product_data=pd.read_excel('Copy of AdventureWorks_Database.xlsx','Product')
```

```
In [12]: Product_data
```

Out[12]:

	ProductKey	ProductName	SubCategory	Category	StandardCost	Color	ListPrice	DaysToManufacture	ProductLine	ModelName	
0	1	Adjustable Race	NaN	NaN	NaN	NaN	NaN	0	NaN	NaN	http://www.avis
1	2	Bearing Ball	NaN	NaN	NaN	NaN	NaN	0	NaN	NaN	http://www.avis
2	3	BB Ball Bearing	NaN	NaN	NaN	NaN	NaN	1	NaN	NaN	http://www.avis
3	4	Headset Ball Bearings	NaN	NaN	NaN	NaN	NaN	0	NaN	NaN	http://www.avis
4	5	Blade	NaN	NaN	NaN	NaN	NaN	1	NaN	NaN	http://www.avis
...
601	602	ML Bottom Bracket	Bottom Brackets	Components	44.9506	NaN	101.24	1	NaN	ML Bottom Bracket	http://www.avis
602	603	HL Bottom Bracket	Bottom Brackets	Components	53.9416	NaN	121.49	1	NaN	HL Bottom Bracket	http://www.avis
603	604	Road-750 Black, 44	Road Bikes	Bikes	343.6496	Black	539.99	4	Road	Road-750	http://www.avis
604	605	Road-750 Black, 48	Road Bikes	Bikes	343.6496	Black	539.99	4	Road	Road-750	http://www.avis
605	606	Road-750 Black, 52	Road Bikes	Bikes	343.6496	Black	539.99	4	Road	Road-750	http://www.avis

606 rows × 13 columns

In [13]: Territory_data=pd.read_excel('Copy of AdventureWorks_Database.xlsx','Territory')

In [14]: Territory_data

Out[14]:

	SalesTerritoryKey	Region	Country	Group	RegionImage
0	1	Northwest	United States	North America	http://www.avising.com/me/LearnPBI/DataSources...
1	2	Northeast	United States	North America	http://www.avising.com/me/LearnPBI/DataSources...
2	3	Central	United States	North America	http://www.avising.com/me/LearnPBI/DataSources...
3	4	Southwest	United States	North America	http://www.avising.com/me/LearnPBI/DataSources...
4	5	Southeast	United States	North America	http://www.avising.com/me/LearnPBI/DataSources...
5	6	Canada	Canada	North America	http://www.avising.com/me/LearnPBI/DataSources...
6	7	France	France	Europe	http://www.avising.com/me/LearnPBI/DataSources...
7	8	Germany	Germany	Europe	http://www.avising.com/me/LearnPBI/DataSources...
8	9	Australia	Australia	Pacific	http://www.avising.com/me/LearnPBI/DataSources...
9	10	United Kingdom	United Kingdom	Europe	http://www.avising.com/me/LearnPBI/DataSources...
10	11	NaN	NaN	NaN	http://www.avising.com/me/LearnPBI/DataSources...

In [15]: Sales_data=pd.read_excel('Copy of AdventureWorks_Database.xlsx','Sales')

In [16]: Sales_data

Out[16]:

	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	UnitPrice
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	346	2014-01-01	2014-01-08	28389	1	7	SO43698	1	2	169
2	346	2014-01-01	2014-01-08	25863	1	1	SO43699	1	2	169
3	336	2014-01-01	2014-01-08	14501	1	4	SO43700	1	2	34
4	346	2014-01-01	2014-01-08	11003	1	9	SO43701	1	2	169
...
58184	561	2016-12-30	2017-01-07	13650	1	9	SO74145	1	1	238
58185	584	2016-12-30	2017-01-07	26916	1	9	SO74146	1	1	53
58186	605	2016-12-30	2017-01-07	27473	1	9	SO74147	1	1	53
58187	538	2016-12-30	2017-01-07	27473	1	9	SO74147	2	1	2
58188	490	2016-12-30	2017-01-07	27473	1	9	SO74147	3	1	5

58189 rows × 25 columns



```
In [17]: # Merging Data
temp_data = pd.merge(Sales_data, Product_data, on='ProductKey', how='inner')
```

```
In [18]: temp_data
```


Out[18]:

	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	Un
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	310	2014-01-02	2014-01-09	16624	1	9	SO43703	1	4	89
2	310	2014-01-05	2014-01-12	27601	1	4	SO43713	1	1	357
3	310	2014-01-06	2014-01-13	13590	1	10	SO43721	1	1	357
4	310	2014-01-10	2014-01-17	16522	1	9	SO43735	1	1	357
...
58184	567	2016-12-13	2016-12-20	15662	1	10	SO72823	1	2	37
58185	567	2016-12-14	2016-12-21	26646	1	4	SO72925	1	4	18
58186	567	2016-12-20	2016-12-27	11363	1	9	SO73411	1	2	37
58187	567	2016-12-22	2016-12-29	11944	1	9	SO73577	1	1	74
58188	567	2016-12-25	2017-01-02	11261	1	4	SO73768	1	1	74

58189 rows × 37 columns

```
In [19]: df2 = pd.merge(temp_data, Customer_data, on='CustomerKey', how='inner')
df2 = pd.merge(df2, Territory_data, on='SalesTerritoryKey', how='inner')
```

```
In [20]: df2
```

```
Out[20]:
```

	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	Un
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	600	2016-04-16	2016-04-23	21768	1	6	SO56212	1	1	55
2	310	2014-01-30	2014-02-06	21727	1	6	SO43833	1	4	89
3	479	2016-11-29	2016-12-05	21727	1	6	SO71614	2	1	
4	477	2016-11-29	2016-12-05	21727	1	6	SO71614	3	1	
...
58184	528	2016-11-07	2016-11-14	13145	1	2	SO70064	2	1	
58185	361	2016-11-07	2016-11-14	13145	1	2	SO70064	1	1	229
58186	480	2016-11-07	2016-11-14	13145	1	2	SO70064	4	1	
58187	530	2016-02-06	2016-02-13	27040	1	2	SO52124	1	1	
58188	480	2016-02-06	2016-02-13	27040	2	2	SO52124	2	1	

58189 rows × 57 columns

```
In [25]: df2.info()
```

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

```
RangeIndex: 58189 entries, 0 to 58188
```

```
Data columns (total 57 columns):
```

#	Column	Non-Null Count	Dtype
0	ProductKey	58189 non-null	int64
1	OrderDate	58189 non-null	datetime64[ns]
2	ShipDate	58189 non-null	datetime64[ns]
3	CustomerKey	58189 non-null	int64
4	PromotionKey	58189 non-null	int64
5	SalesTerritoryKey	58189 non-null	int64
6	SalesOrderNumber	58189 non-null	object
7	SalesOrderLineNumber	58189 non-null	int64
8	OrderQuantity	58189 non-null	int64
9	UnitPrice	58189 non-null	float64
10	TotalProductCost	58189 non-null	float64
11	SalesAmount	58189 non-null	float64
12	TaxAmt	58189 non-null	float64
13	Unnamed: 13	0 non-null	float64
14	Unnamed: 14	0 non-null	float64
15	Unnamed: 15	58189 non-null	float64
16	Unnamed: 16	58189 non-null	float64
17	Unnamed: 17	0 non-null	float64
18	Unnamed: 18	58189 non-null	float64
19	Unnamed: 19	0 non-null	float64
20	StandardCost_x	58189 non-null	float64
21	List Price	58189 non-null	float64
22	Unnamed: 22	0 non-null	float64
23	diif std cost	58189 non-null	int64
24	diff list price	58189 non-null	int64
25	ProductName	58189 non-null	object
26	SubCategory	58189 non-null	object
27	Category	58189 non-null	object
28	StandardCost_y	58189 non-null	float64
29	Color	30747 non-null	object
30	ListPrice	58189 non-null	float64
31	DaysToManufacture	58189 non-null	int64
32	ProductLine	58189 non-null	object
33	ModelName	58189 non-null	object
34	Photo	58189 non-null	object
35	ProductDescription	58189 non-null	object
36	StartDate	58189 non-null	datetime64[ns]
37	FirstName	58189 non-null	object
38	LastName	58189 non-null	object

```
39  FullName          58189 non-null object
40  BirthDate         58189 non-null datetime64[ns]
41  MaritalStatus     58189 non-null object
42  Gender            58189 non-null object
43  YearlyIncome       58189 non-null int64
44  TotalChildren     58189 non-null int64
45  NumberChildrenAtHome 58189 non-null int64
46  Education         58189 non-null object
47  Occupation        58189 non-null object
48  HouseOwnerFlag    58189 non-null int64
49  NumberCarsOwned   58189 non-null int64
50  AddressLine1      58189 non-null object
51  DateFirstPurchase 58189 non-null datetime64[ns]
52  CommuteDistance   58189 non-null object
53  Region            58189 non-null object
54  Country           58189 non-null object
55  Group             58189 non-null object
56  RegionImage       58189 non-null object
dtypes: datetime64[ns](5), float64(16), int64(14), object(22)
memory usage: 25.3+ MB
```

```
In [26]: df2.duplicated().sum()
```

```
Out[26]: 0
```

```
In [27]: df3 = pd.merge(df, df2, how='left')
```

```
In [28]: df3
```

Out[28]:

	Category	Subcategory	ProductName	ProductKey	Jan, 2016	Feb, 2016	Mar, 2016	Apr, 2016	May, 2016	Jun, 2016	...	Occupation	HouseOwnerFlag	Nu
0	Accessories	Bike Racks	Hitch Rack - 4-Bike	483.0	1131	2635	4134	2179	2637	3279	...	Skilled Manual	1.0	
1	Accessories	Bike Racks	Hitch Rack - 4-Bike	483.0	1131	2635	4134	2179	2637	3279	...	Skilled Manual	1.0	
2	Accessories	Bike Racks	Hitch Rack - 4-Bike	483.0	1131	2635	4134	2179	2637	3279	...	Clerical	1.0	
3	Accessories	Bike Racks	Hitch Rack - 4-Bike	483.0	1131	2635	4134	2179	2637	3279	...	Professional	1.0	
4	Accessories	Bike Racks	Hitch Rack - 4-Bike	483.0	1131	2635	4134	2179	2637	3279	...	Skilled Manual	1.0	
...
11854	Clothing	Vests	Classic Vest, S	471.0	980	2008	1980	2312	2763	2591	...	Manual	0.0	
11855	Clothing	Vests	Classic Vest, S	471.0	980	2008	1980	2312	2763	2591	...	Clerical	1.0	
11856	Clothing	Vests	Classic Vest, S	471.0	980	2008	1980	2312	2763	2591	...	Manual	1.0	
11857	SubTotal Clothing	NaN	NaN	NaN	6589	23652	27004	24787	25773	33091	...	NaN	NaN	
11858	Grand Total	NaN	NaN	NaN	871119	858350	999795	1131721	1226351	1713430	...	NaN	NaN	

11859 rows × 71 columns



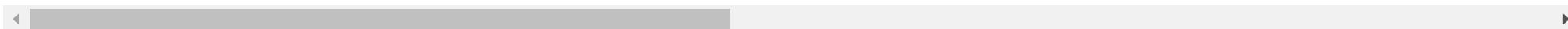
In [29]:

```
df2
```

Out[29]:

	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	Un
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	600	2016-04-16	2016-04-23	21768	1	6	SO56212	1	1	53
2	310	2014-01-30	2014-02-06	21727	1	6	SO43833	1	4	89
3	479	2016-11-29	2016-12-05	21727	1	6	SO71614	2	1	
4	477	2016-11-29	2016-12-05	21727	1	6	SO71614	3	1	
...
58184	528	2016-11-07	2016-11-14	13145	1	2	SO70064	2	1	
58185	361	2016-11-07	2016-11-14	13145	1	2	SO70064	1	1	229
58186	480	2016-11-07	2016-11-14	13145	1	2	SO70064	4	1	
58187	530	2016-02-06	2016-02-13	27040	1	2	SO52124	1	1	
58188	480	2016-02-06	2016-02-13	27040	2	2	SO52124	2	1	

58189 rows × 57 columns



In [30]:

```
# Assessing Data
df2.info()
```

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

```
RangeIndex: 58189 entries, 0 to 58188
```

```
Data columns (total 57 columns):
```

#	Column	Non-Null Count	Dtype
0	ProductKey	58189 non-null	int64
1	OrderDate	58189 non-null	datetime64[ns]
2	ShipDate	58189 non-null	datetime64[ns]
3	CustomerKey	58189 non-null	int64
4	PromotionKey	58189 non-null	int64
5	SalesTerritoryKey	58189 non-null	int64
6	SalesOrderNumber	58189 non-null	object
7	SalesOrderLineNumber	58189 non-null	int64
8	OrderQuantity	58189 non-null	int64
9	UnitPrice	58189 non-null	float64
10	TotalProductCost	58189 non-null	float64
11	SalesAmount	58189 non-null	float64
12	TaxAmt	58189 non-null	float64
13	Unnamed: 13	0 non-null	float64
14	Unnamed: 14	0 non-null	float64
15	Unnamed: 15	58189 non-null	float64
16	Unnamed: 16	58189 non-null	float64
17	Unnamed: 17	0 non-null	float64
18	Unnamed: 18	58189 non-null	float64
19	Unnamed: 19	0 non-null	float64
20	StandardCost_x	58189 non-null	float64
21	List Price	58189 non-null	float64
22	Unnamed: 22	0 non-null	float64
23	diif std cost	58189 non-null	int64
24	diff list price	58189 non-null	int64
25	ProductName	58189 non-null	object
26	SubCategory	58189 non-null	object
27	Category	58189 non-null	object
28	StandardCost_y	58189 non-null	float64
29	Color	30747 non-null	object
30	ListPrice	58189 non-null	float64
31	DaysToManufacture	58189 non-null	int64
32	ProductLine	58189 non-null	object
33	ModelName	58189 non-null	object
34	Photo	58189 non-null	object
35	ProductDescription	58189 non-null	object
36	StartDate	58189 non-null	datetime64[ns]
37	FirstName	58189 non-null	object
38	LastName	58189 non-null	object


```
39  FullName          58189 non-null object
40  BirthDate         58189 non-null datetime64[ns]
41  MaritalStatus     58189 non-null object
42  Gender            58189 non-null object
43  YearlyIncome       58189 non-null int64
44  TotalChildren     58189 non-null int64
45  NumberChildrenAtHome 58189 non-null int64
46  Education         58189 non-null object
47  Occupation        58189 non-null object
48  HouseOwnerFlag    58189 non-null int64
49  NumberCarsOwned   58189 non-null int64
50  AddressLine1      58189 non-null object
51  DateFirstPurchase 58189 non-null datetime64[ns]
52  CommuteDistance   58189 non-null object
53  Region            58189 non-null object
54  Country            58189 non-null object
55  Group             58189 non-null object
56  RegionImage       58189 non-null object
dtypes: datetime64[ns](5), float64(16), int64(14), object(22)
memory usage: 25.3+ MB
```

```
In [31]: # Check shape of the data after merging
print(f"Number of Rows: {df2.shape[0]}")
print(f"Number of Columns: {df2.shape[1]} \n")
```

```
Number of Rows: 58189
Number of Columns: 57
```

```
In [32]: df2.describe().transpose()
```

Out[32]:

	count	mean	min	25%	50%	75%	max	std
ProductKey	58189.0	437.208304	214.0	358.0	479.0	529.0	606.0	118.099746
OrderDate	58189	2016-06-03 03:56:09.605939200	2014-01-01 00:00:00	2016-04-01 00:00:00	2016-07-07 00:00:00	2016-10-10 00:00:00	2016-12-30 00:00:00	NaN
ShipDate	58189	2016-06-10 04:03:24.657237760	2014-01-08 00:00:00	2016-04-08 00:00:00	2016-07-14 00:00:00	2016-10-17 00:00:00	2017-01-07 00:00:00	NaN
CustomerKey	58189.0	18853.00464	11000.0	14012.0	18151.0	23450.0	29483.0	5433.374315
PromotionKey	58189.0	1.043427	1.0	1.0	1.0	1.0	14.0	0.348948
SalesTerritoryKey	58189.0	6.261716	1.0	4.0	7.0	9.0	10.0	2.960248
SalesOrderLineNumber	58189.0	1.887453	1.0	1.0	2.0	2.0	8.0	1.018829
OrderQuantity	58189.0	1.569386	1.0	1.0	1.0	2.0	4.0	1.047532
UnitPrice	58189.0	413.888218	0.5725	4.99	24.49	269.995	3578.27	833.052938
TotalProductCost	58189.0	296.539185	0.8565	3.3623	12.1924	343.6496	2171.2942	560.171436
SalesAmount	58189.0	503.66627	2.29	8.99	32.6	539.99	3578.27	941.462817
TaxAmt	58189.0	40.293303	0.1832	0.7192	2.608	43.1992	286.2616	75.317027
Unnamed: 13	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
Unnamed: 14	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
Unnamed: 15	58189.0	503.666269	2.29	8.99	32.6	539.99	3578.27	941.462815
Unnamed: 16	58189.0	0.000001	0.0	0.0	0.0	0.0	0.0003	0.000014
Unnamed: 17	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
Unnamed: 18	58189.0	38.398254	-5106.9068	1.4335	6.2537	21.9037	1487.8356	667.349417
Unnamed: 19	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
StandardCost_x	58189.0	296.539185	0.8565	3.3623	12.1924	343.6496	2171.2942	560.171436
List Price	58189.0	503.66627	2.29	8.99	32.6	539.99	3578.27	941.462817
Unnamed: 22	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
diif std cost	58189.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

	count	mean	min	25%	50%	75%	max	std
diff list price	58189.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
StandardCost_y	58189.0	296.539185	0.8565	3.3623	12.1924	343.6496	2171.2942	560.171436
ListPrice	58189.0	503.66627	2.29	8.99	32.6	539.99	3578.27	941.462817
DaysToManufacture	58189.0	1.045215	0.0	0.0	0.0	4.0	4.0	1.757395
StartDate	58189	2007-05-14 02:44:51.848974848	2005-07-01 00:00:00	2007-07-01 00:00:00	2007-07-01 00:00:00	2007-07-01 00:00:00	2007-07-01 00:00:00	NaN
BirthDate	58189	1962-03-02 12:33:19.305710720	1910-08-13 00:00:00	1954-12-20 00:00:00	1963-09-19 00:00:00	1970-07-08 00:00:00	1980-12-26 00:00:00	NaN
YearlyIncome	58189.0	59769.887779	10000.0	30000.0	60000.0	80000.0	170000.0	33128.041818
TotalChildren	58189.0	1.838921	0.0	0.0	2.0	3.0	5.0	1.614467
NumberChildrenAtHome	58189.0	1.073502	0.0	0.0	0.0	2.0	5.0	1.580055
HouseOwnerFlag	58189.0	0.69056	0.0	0.0	1.0	1.0	1.0	0.462267
NumberCarsOwned	58189.0	1.502466	0.0	1.0	2.0	2.0	4.0	1.155496
DateFirstPurchase	58189	2015-12-23 02:50:33.356820224	2014-01-01 00:00:00	2015-06-21 00:00:00	2016-03-12 00:00:00	2016-07-26 00:00:00	2016-12-30 00:00:00	NaN

```
In [33]: # Check for duplicate data
df2.duplicated().sum()
```

```
Out[33]: 0
```

```
In [34]: df2.columns
```

```
Out[34]: Index(['ProductKey', 'OrderDate', 'ShipDate', 'CustomerKey', 'PromotionKey',  
            'SalesTerritoryKey', 'SalesOrderNumber', 'SalesOrderLineNumber',  
            'OrderQuantity', 'UnitPrice', 'TotalProductCost', 'SalesAmount',  
            'TaxAmt', 'Unnamed: 13', 'Unnamed: 14', 'Unnamed: 15', 'Unnamed: 16',  
            'Unnamed: 17', 'Unnamed: 18', 'Unnamed: 19', 'StandardCost_x',  
            'List Price', 'Unnamed: 22', 'diif std cost', 'diff list price',  
            'ProductName', 'SubCategory', 'Category', 'StandardCost_y', 'Color',  
            'ListPrice', 'DaysToManufacture', 'ProductLine', 'ModelName', 'Photo',  
            'ProductDescription', 'StartDate', 'FirstName', 'LastName', 'FullName',  
            'BirthDate', 'MaritalStatus', 'Gender', 'YearlyIncome', 'TotalChildren',  
            'NumberChildrenAtHome', 'Education', 'Occupation', 'HouseOwnerFlag',  
            'NumberCarsOwned', 'AddressLine1', 'DateFirstPurchase',  
            'CommuteDistance', 'Region', 'Country', 'Group', 'RegionImage'],  
            dtype='object')
```

```
In [35]: df2.drop('Unnamed: 13',axis=1,inplace=True)
```

```
In [36]: df2.columns
```

```
Out[36]: Index(['ProductKey', 'OrderDate', 'ShipDate', 'CustomerKey', 'PromotionKey',  
            'SalesTerritoryKey', 'SalesOrderNumber', 'SalesOrderLineNumber',  
            'OrderQuantity', 'UnitPrice', 'TotalProductCost', 'SalesAmount',  
            'TaxAmt', 'Unnamed: 14', 'Unnamed: 15', 'Unnamed: 16', 'Unnamed: 17',  
            'Unnamed: 18', 'Unnamed: 19', 'StandardCost_x', 'List Price',  
            'Unnamed: 22', 'diif std cost', 'diff list price', 'ProductName',  
            'SubCategory', 'Category', 'StandardCost_y', 'Color', 'ListPrice',  
            'DaysToManufacture', 'ProductLine', 'ModelName', 'Photo',  
            'ProductDescription', 'StartDate', 'FirstName', 'LastName', 'FullName',  
            'BirthDate', 'MaritalStatus', 'Gender', 'YearlyIncome', 'TotalChildren',  
            'NumberChildrenAtHome', 'Education', 'Occupation', 'HouseOwnerFlag',  
            'NumberCarsOwned', 'AddressLine1', 'DateFirstPurchase',  
            'CommuteDistance', 'Region', 'Country', 'Group', 'RegionImage'],  
            dtype='object')
```

```
In [37]: df2.drop(df2.columns[[13,14,15,16,17,18,21]],axis = 1,inplace=True)
```

```
In [38]: df2
```

Out[38]:

	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	Un
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	600	2016-04-16	2016-04-23	21768	1	6	SO56212	1	1	53
2	310	2014-01-30	2014-02-06	21727	1	6	SO43833	1	4	89
3	479	2016-11-29	2016-12-05	21727	1	6	SO71614	2	1	
4	477	2016-11-29	2016-12-05	21727	1	6	SO71614	3	1	
...
58184	528	2016-11-07	2016-11-14	13145	1	2	SO70064	2	1	
58185	361	2016-11-07	2016-11-14	13145	1	2	SO70064	1	1	229
58186	480	2016-11-07	2016-11-14	13145	1	2	SO70064	4	1	
58187	530	2016-02-06	2016-02-13	27040	1	2	SO52124	1	1	
58188	480	2016-02-06	2016-02-13	27040	2	2	SO52124	2	1	

58189 rows × 49 columns

```
In [39]: # Handling missing data
df2.isnull().sum()
```

```
Out[39]: ProductKey          0
         OrderDate          0
         ShipDate           0
         CustomerKey        0
         PromotionKey       0
         SalesTerritoryKey  0
         SalesOrderNumber   0
         SalesOrderLineNumber 0
         OrderQuantity       0
         UnitPrice           0
         TotalProductCost    0
         SalesAmount         0
         TaxAmt              0
         StandardCost_x     0
         List Price          0
         diif std cost       0
         diff list price     0
         ProductName         0
         SubCategory         0
         Category            0
         StandardCost_y     0
         Color               27442
         ListPrice           0
         DaysToManufacture   0
         ProductLine         0
         ModelName           0
         Photo               0
         ProductDescription  0
         StartDate           0
         FirstName           0
         LastName            0
         FullName            0
         BirthDate           0
         MaritalStatus       0
         Gender              0
         YearlyIncome         0
         TotalChildren        0
         NumberChildrenAtHome 0
         Education           0
         Occupation           0
         HouseOwnerFlag      0
         NumberCarsOwned     0
         AddressLine1        0
         DateFirstPurchase   0
```

CommuteDistance	0
Region	0
Country	0
Group	0
RegionImage	0
dtype: int64	

In [40]: df2.dtypes

```
Out[40]: ProductKey          int64
OrderDate      datetime64[ns]
ShipDate       datetime64[ns]
CustomerKey    int64
PromotionKey   int64
SalesTerritoryKey int64
SalesOrderNumber object
SalesOrderLineNumber int64
OrderQuantity  int64
UnitPrice      float64
TotalProductCost float64
SalesAmount    float64
TaxAmt         float64
StandardCost_x float64
List Price     float64
diif std cost  int64
diff list price int64
ProductName    object
SubCategory    object
Category       object
StandardCost_y float64
Color          object
ListPrice      float64
DaysToManufacture int64
ProductLine    object
ModelName      object
Photo          object
ProductDescription object
StartDate      datetime64[ns]
FirstName      object
LastName       object
FullName       object
BirthDate      datetime64[ns]
MaritalStatus  object
Gender         object
YearlyIncome   int64
TotalChildren  int64
NumberChildrenAtHome int64
Education      object
Occupation     object
HouseOwnerFlag int64
NumberCarsOwned int64
AddressLine1   object
DateFirstPurchase datetime64[ns]
```



```
CommuteDistance      object
Region               object
Country              object
Group                object
RegionImage          object
dtype: object
```

```
In [41]: def missing_pct(df2):
# Calculate missing value and their percentage for each column
missing_count_percent = df2.isnull().sum() * 100 / df2.shape[0]
df2_missing_count_percent = pd.DataFrame(missing_count_percent).round(2)
df2_missing_count_percent = df2_missing_count_percent.reset_index().rename(
    columns={
        'index': 'Column',
        0: 'Missing_Percentage (%)'
    }
)

df2_missing_value = df2.isnull().sum()
df2_missing_value = df2_missing_value.reset_index().rename(
    columns={
        'index': 'Column',
        0: 'Missing_value_count'
    }
)

# Sort the data frame
#df2_missing = df2_missing.sort_values('Missing_Percentage (%)', ascending=False)
Final = df2_missing_value.merge(df2_missing_count_percent, how = 'inner', left_on = 'Column', right_on = 'Column')
Final = Final.sort_values(by = 'Missing_Percentage (%)', ascending = False)
return Final
```

```
In [42]: # Applying the custom function
missing_pct(df2)
```

Out[42]:

	Column	Missing_value_count	Missing_Percentage (%)
21	Color	27442	47.16
0	ProductKey	0	0.00
36	TotalChildren	0	0.00
27	ProductDescription	0	0.00
28	StartDate	0	0.00
29	FirstName	0	0.00
30	LastName	0	0.00
31	FullName	0	0.00
32	BirthDate	0	0.00
33	MaritalStatus	0	0.00
34	Gender	0	0.00
35	YearlyIncome	0	0.00
37	NumberChildrenAtHome	0	0.00
25	ModelName	0	0.00
38	Education	0	0.00
39	Occupation	0	0.00
40	HouseOwnerFlag	0	0.00
41	NumberCarsOwned	0	0.00
42	AddressLine1	0	0.00
43	DateFirstPurchase	0	0.00
44	CommuteDistance	0	0.00
45	Region	0	0.00
46	Country	0	0.00
47	Group	0	0.00

	Column	Missing_value_count	Missing_Percentage (%)
26	Photo	0	0.00
24	ProductLine	0	0.00
1	OrderDate	0	0.00
11	SalesAmount	0	0.00
2	ShipDate	0	0.00
3	CustomerKey	0	0.00
4	PromotionKey	0	0.00
5	SalesTerritoryKey	0	0.00
6	SalesOrderNumber	0	0.00
7	SalesOrderLineNumber	0	0.00
8	OrderQuantity	0	0.00
9	UnitPrice	0	0.00
10	TotalProductCost	0	0.00
12	TaxAmt	0	0.00
23	DaysToManufacture	0	0.00
13	StandardCost_x	0	0.00
14	List Price	0	0.00
15	diif std cost	0	0.00
16	diff list price	0	0.00
17	ProductName	0	0.00
18	SubCategory	0	0.00
19	Category	0	0.00
20	StandardCost_y	0	0.00
22	ListPrice	0	0.00

	Column	Missing_value_count	Missing_Percentage (%)
48	RegionImage	0	0.00

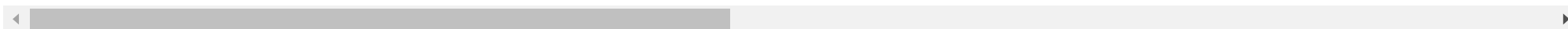
```
In [43]: # Drop columns with nan values  
df2= df2.dropna(axis=1)
```

```
In [44]: df2
```

Out[44]:

	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	Un
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	600	2016-04-16	2016-04-23	21768	1	6	SO56212	1	1	53
2	310	2014-01-30	2014-02-06	21727	1	6	SO43833	1	4	89
3	479	2016-11-29	2016-12-05	21727	1	6	SO71614	2	1	
4	477	2016-11-29	2016-12-05	21727	1	6	SO71614	3	1	
...
58184	528	2016-11-07	2016-11-14	13145	1	2	SO70064	2	1	
58185	361	2016-11-07	2016-11-14	13145	1	2	SO70064	1	1	229
58186	480	2016-11-07	2016-11-14	13145	1	2	SO70064	4	1	
58187	530	2016-02-06	2016-02-13	27040	1	2	SO52124	1	1	
58188	480	2016-02-06	2016-02-13	27040	2	2	SO52124	2	1	

58189 rows × 48 columns



In [45]: df2.columns

```
Out[45]: Index(['ProductKey', 'OrderDate', 'ShipDate', 'CustomerKey', 'PromotionKey',  
        'SalesTerritoryKey', 'SalesOrderNumber', 'SalesOrderLineNumber',  
        'OrderQuantity', 'UnitPrice', 'TotalProductCost', 'SalesAmount',  
        'TaxAmt', 'StandardCost_x', 'List Price', 'diif std cost',  
        'diff list price', 'ProductName', 'SubCategory', 'Category',  
        'StandardCost_y', 'ListPrice', 'DaysToManufacture', 'ProductLine',  
        'ModelName', 'Photo', 'ProductDescription', 'StartDate', 'FirstName',  
        'LastName', 'FullName', 'BirthDate', 'MaritalStatus', 'Gender',  
        'YearlyIncome', 'TotalChildren', 'NumberChildrenAtHome', 'Education',  
        'Occupation', 'HouseOwnerFlag', 'NumberCarsOwned', 'AddressLine1',  
        'DateFirstPurchase', 'CommuteDistance', 'Region', 'Country', 'Group',  
        'RegionImage'],  
        dtype='object')
```

Adding Columns

```
In [46]: # Extracting Year from OrderDate  
df2['sale_year'] = df2['OrderDate'].dt.year  
  
# Extracting Month from OrderDate  
df2['sale_month'] = df2['OrderDate'].dt.month  
  
# Extracting day from OrderDate  
df2['sale_day'] = df2['OrderDate'].dt.day  
  
# Extracting dayofweek from OrderDate  
df2['sale_week'] = df2['OrderDate'].dt.dayofweek  
  
# Extracting day_name from OrderDate  
df2['sale_day_name'] = df2['OrderDate'].dt.day_name()  
  
# Extracting Month Year from OrderDate  
df2['year_month'] = df2['OrderDate'].apply(lambda x:x.strftime('%Y-%m'))  
  
# Calculate Total Invoice Amount  
df2['total_invoice_amount'] = df2['SalesAmount'] + df2['TaxAmt']  
  
# Considering only salesamount and total_sales_amount to calculate profit  
df2['profit'] = (df2['UnitPrice']*df2['OrderQuantity']) - df2['TotalProductCost']  
  
# Removing extra character from the string  
df2['ProductName'] = df2['ProductName'].str.replace(',', '-')
```

```
# Calculate Age
df2['Age'] = df2['OrderDate'].dt.year - df2['BirthDate'].dt.year
```

In [47]: df2

Out[47]:

	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	UnitPrice
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	600	2016-04-16	2016-04-23	21768	1	6	SO56212	1	1	53
2	310	2014-01-30	2014-02-06	21727	1	6	SO43833	1	4	89
3	479	2016-11-29	2016-12-05	21727	1	6	SO71614	2	1	
4	477	2016-11-29	2016-12-05	21727	1	6	SO71614	3	1	
...
58184	528	2016-11-07	2016-11-14	13145	1	2	SO70064	2	1	
58185	361	2016-11-07	2016-11-14	13145	1	2	SO70064	1	1	229
58186	480	2016-11-07	2016-11-14	13145	1	2	SO70064	4	1	
58187	530	2016-02-06	2016-02-13	27040	1	2	SO52124	1	1	
58188	480	2016-02-06	2016-02-13	27040	2	2	SO52124	2	1	

58189 rows × 11 columns



Exploring data

List of product's category

```
In [48]: df2.groupby('sale_year')['SalesAmount'].sum()
```

```
Out[48]: sale_year
2014      7.072084e+06
2015      5.762134e+06
2016      1.647362e+07
Name: SalesAmount, dtype: float64
```

List of product's subcategory

```
In [49]: df2['SubCategory'].unique().tolist()
```

```
Out[49]: ['Road Bikes',
'Mountain Bikes',
'Bottles and Cages',
'Gloves',
'Tires and Tubes',
'Helmets',
'Touring Bikes',
'Jerseys',
'Cleaners',
'Caps',
'Hydration Packs',
'Socks',
'Fenders',
'Vests',
'Bike Racks',
'Bike Stands',
'Shorts']
```

Analysing UnitPrice

```
In [50]: Avg_unit_price = df2.groupby('ProductKey')['UnitPrice'].mean()
ax = sns.distplot(Avg_unit_price, kde=True, hist=True, color='#374045')
ax.set(title='Distribution of Average unit price',
        xlabel='Average Unit Price')
```



```
C:\Users\ersum\AppData\Local\Temp\ipykernel_15792\1028881447.py:2: UserWarning:
```

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

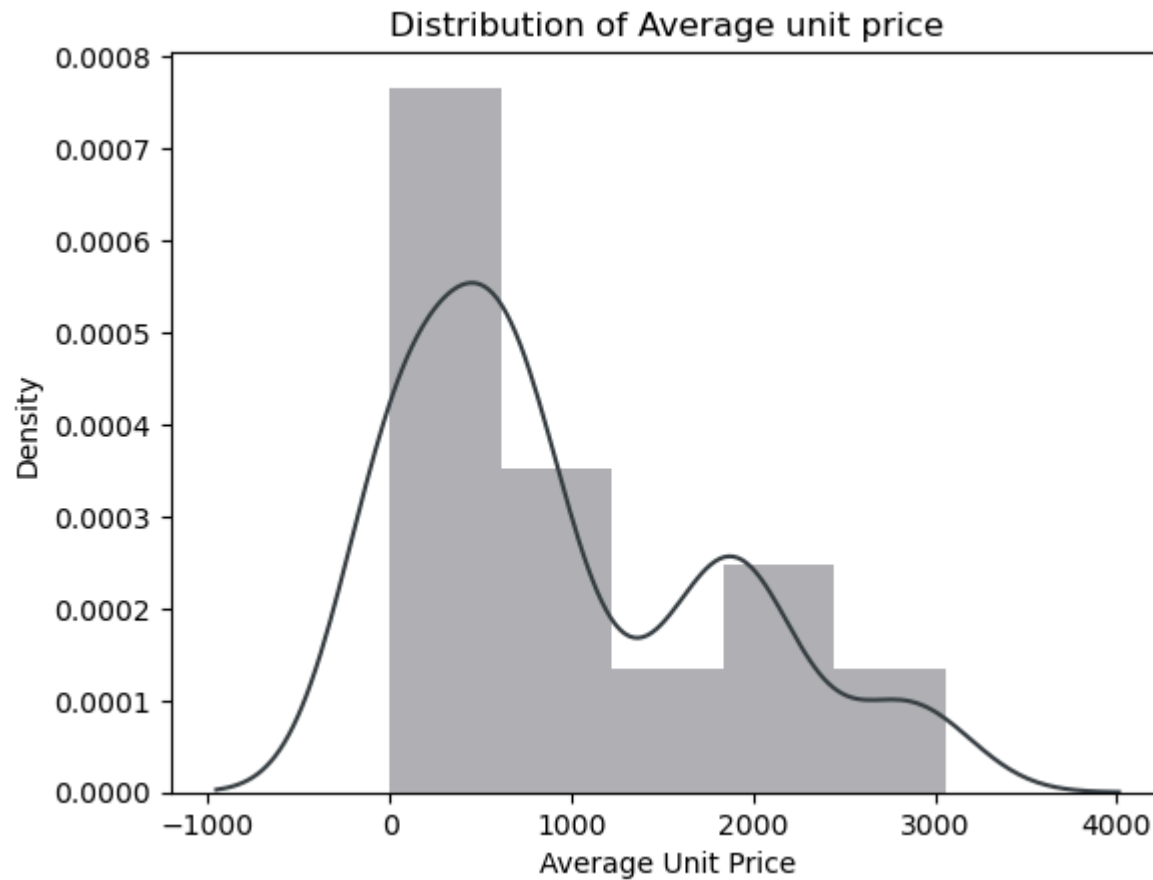
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

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

```
ax = sns.distplot(Avg_unit_price, kde=True, hist=True, color='#374045')
```

Out[50]:

```
[Text(0.5, 1.0, 'Distribution of Average unit price'),  
Text(0.5, 0, 'Average Unit Price')]
```



```
In [51]: # Maximum of the product unit price is below $1000
```

Sales order number distribution

```
In [52]: n_orders = df2.groupby(['CustomerKey'])['SalesOrderNumber'].nunique()
multi_orders_perc = np.sum(n_orders > 1)/df2['CustomerKey'].nunique()
print(f"{100*multi_orders_perc:.2f}% of customers ordered more than once.")
```

36.97% of customers ordered more than once.

```
In [53]: ax = sns.distplot(n_orders, kde=False, color='#374045')
ax.set(title='Distribution of number of orders per customer',
       xlabel='# of orders',
       ylabel='# of customers');
```

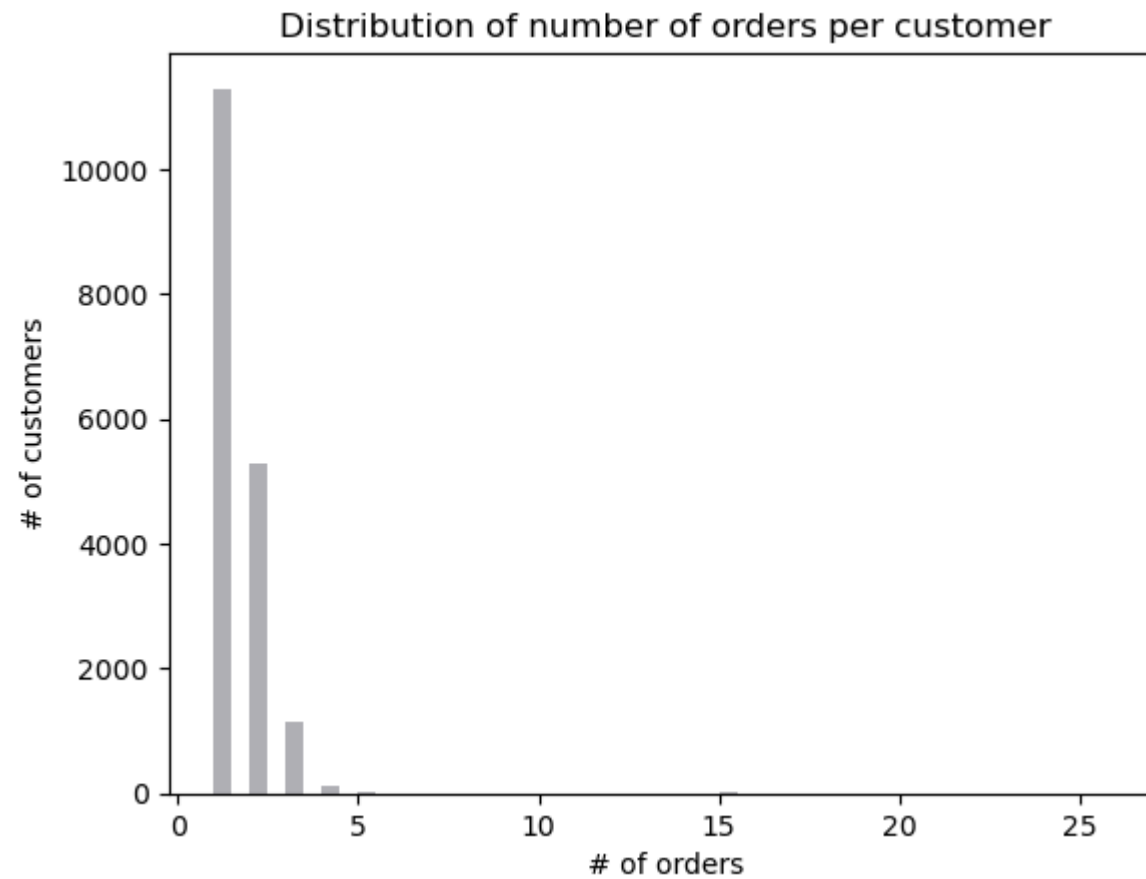
C:\Users\ersum\AppData\Local\Temp\ipykernel_15792\1110922211.py:1: UserWarning:

`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>

```
ax = sns.distplot(n_orders, kde=False, color='#374045')
```



Sales order line number distribution

```
In [54]: n_salesordernumber = df2.groupby(['SalesOrderNumber'])['SalesOrderLineNumber'].transform('max')
ax = sns.distplot(n_salesordernumber, kde=False, color='#374045')
ax.set(title='Distribution of sales order line number',
       xlabel='# of Sales order line number',
       ylabel='# of orders');
```

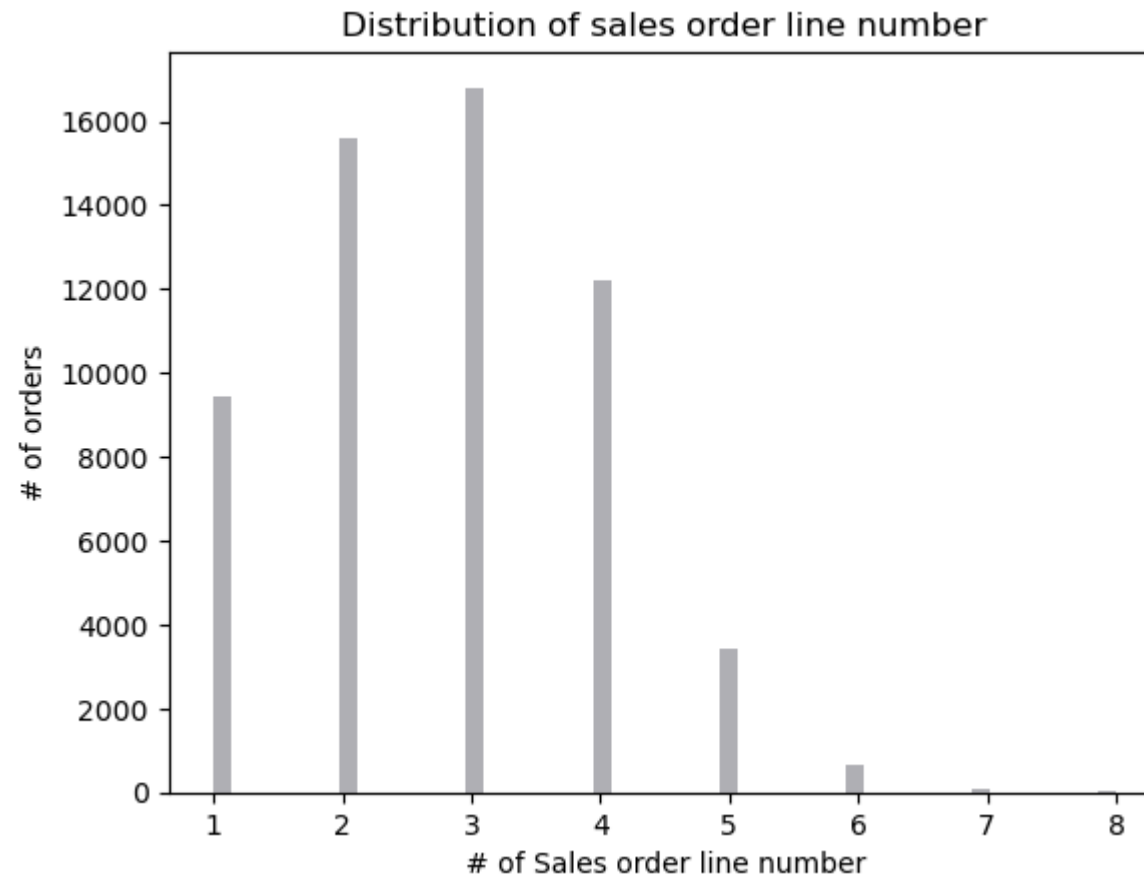
```
C:\Users\ersum\AppData\Local\Temp\ipykernel_15792\1353084701.py:2: UserWarning:
```

```
`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>

```
ax = sns.distplot(n_salesordernumber, kde=False, color='#374045')
```



In [55]:

```
# Most of the time two to three products are ordered in a single order
```

Sales Order Quantity distribution

```
In [56]: n_order_quantity = df2.groupby(['SalesOrderNumber'])['OrderQuantity'].sum()  
ax = sns.distplot(n_order_quantity, kde=True, hist=True,color='#374045')  
ax.set(title='Distribution of order_quantity',  
        xlabel='# of order_quantity',  
        );
```

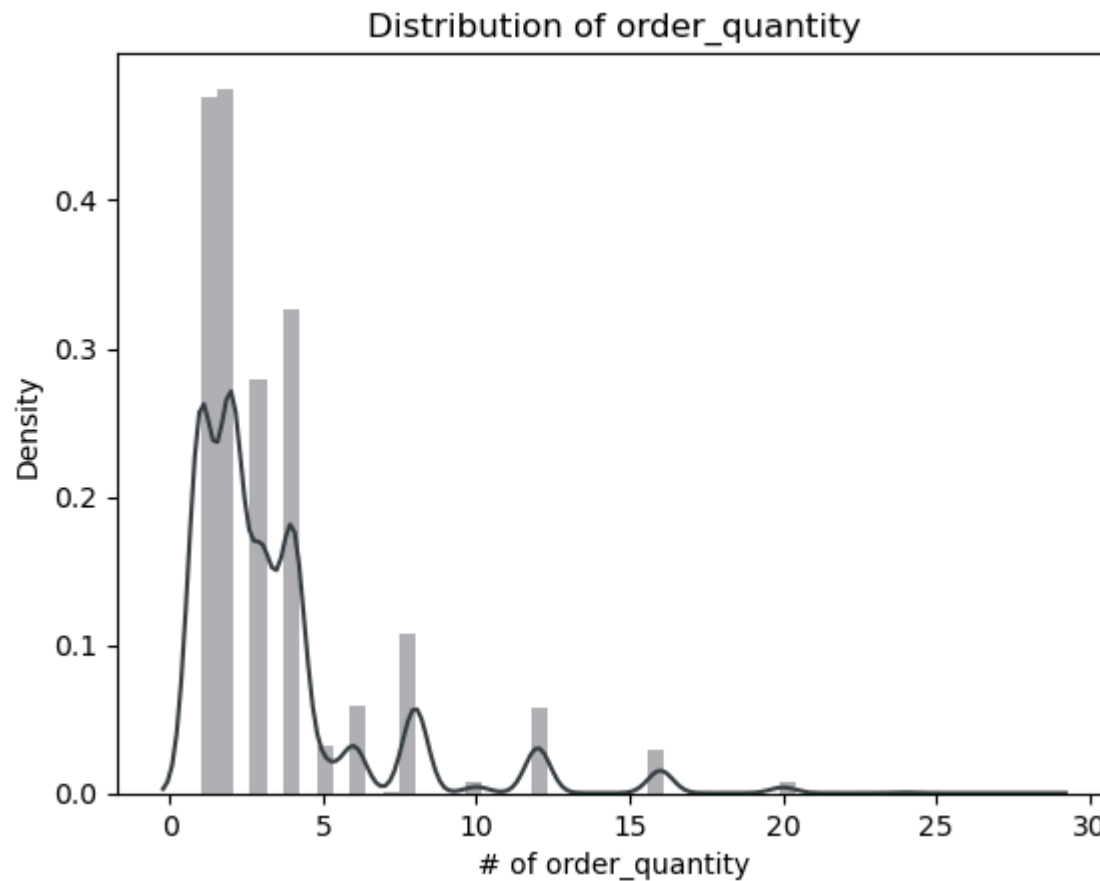
C:\Users\ersum\AppData\Local\Temp\ipykernel_15792\1426996600.py:2: UserWarning:

`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>

```
ax = sns.distplot(n_order_quantity, kde=True, hist=True,color='#374045')
```



```
In [57]: # maximum quantity ordered for a product is below 2.5
```

Age Distribution

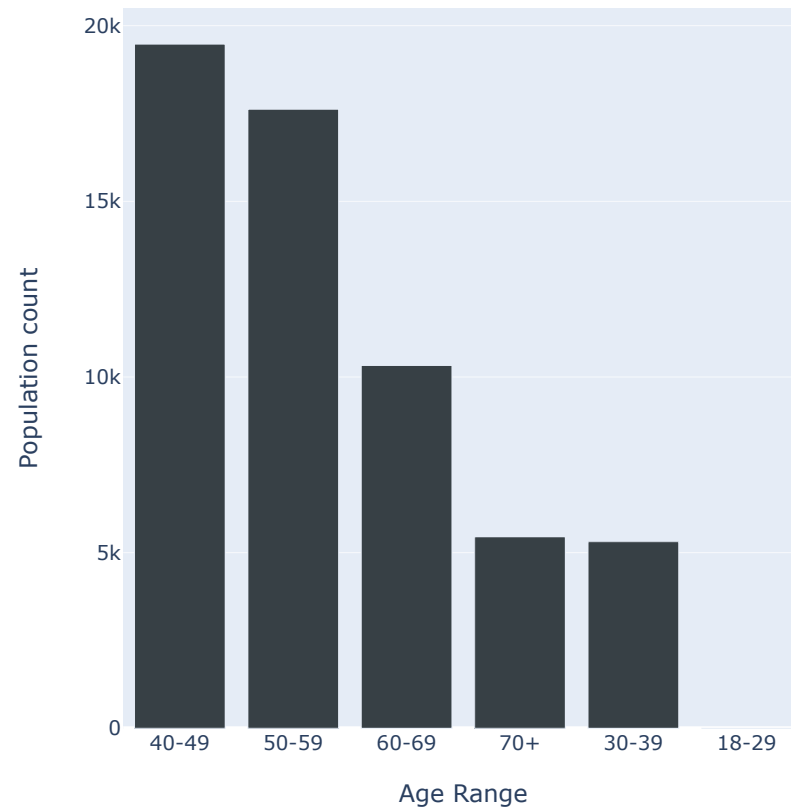
```
In [58]: bins = [18, 30, 40, 50, 60, 70, 120]
labels = ['18-29', '30-39', '40-49', '50-59', '60-69', '70+']
df2['agerange'] = pd.cut(df2.Age, bins, labels = labels, include_lowest = True)

age_distribution = df2['agerange'].value_counts().to_frame().reset_index()

age_distribution.columns = ['Age Range', 'Population count']

fig = px.bar(age_distribution, x='Age Range', y='Population count', color_discrete_sequence=['#374045'])
fig.update_layout()
```

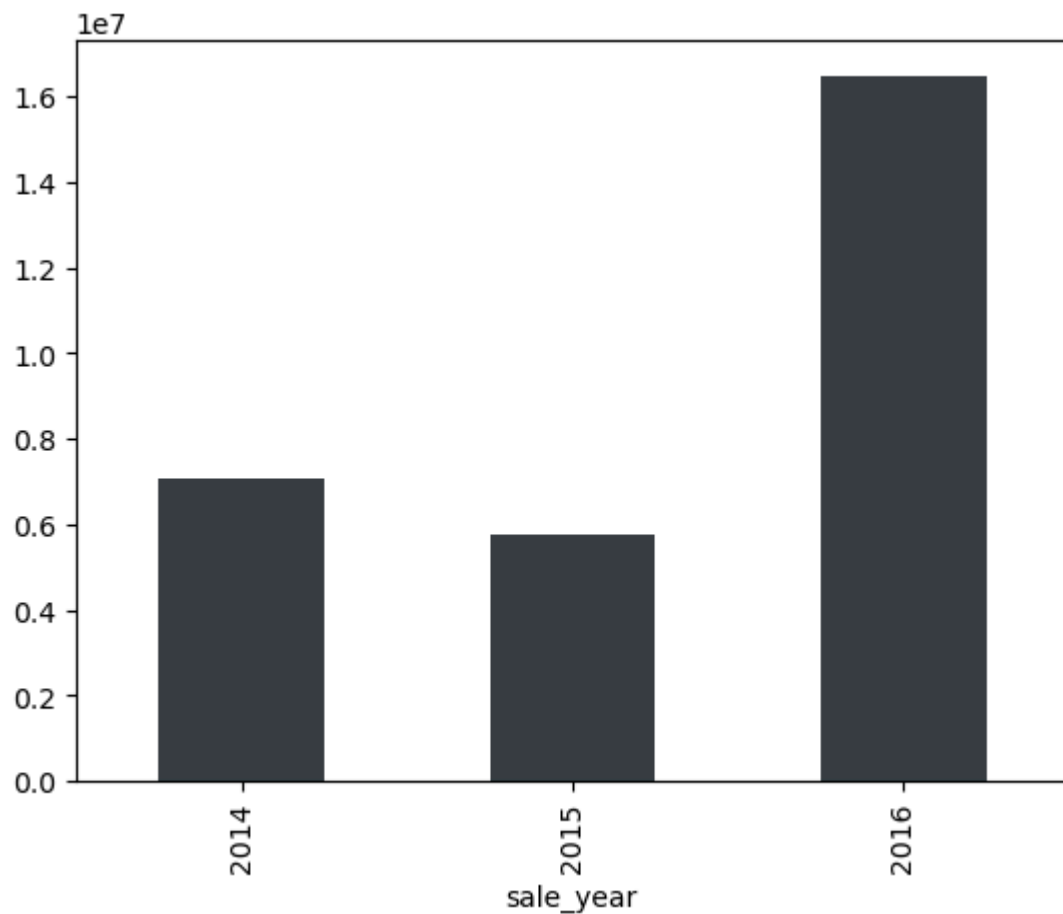
```
autosize=True,  
width=500,  
height=500,  
font=dict(size=10))  
fig.show()
```



In [59]: *# The maximum no. of clients are between the ages of 40 and 59.*

Sales Year wise sales

```
In [60]: df2.groupby('sale_year')['SalesAmount'].sum().plot(kind='bar', color='#374045');
```



```
In [62]: # sales in 2016 are maximum.sales initially start decreasing from 2014 to 2015 the increases gradually in 2016.
```

Top 5 Selling Product

```
In [64]: top_selling_product = df2.groupby(['Category', 'SubCategory', 'ProductName'])['OrderQuantity'].sum().sort_values(ascending=False)
top_selling_product.head(5)
```



```
Out[64]:
```

Category	SubCategory	ProductName	
Accessories	Bottles and Cages	Water Bottle - 30 oz.	6370
	Tires and Tubes	Patch Kit/8 Patches	4705
		Mountain Tire Tube	4551
		Road Tire Tube	3544
	Helmets	Sport-100 Helmet- Red	3398

Name: OrderQuantity, dtype: int64

Quantity ordered based on category and subcategory from 2014 to 2016

```
In [65]: cat_subcat_qty = df2.groupby(['sale_year', 'Category', 'SubCategory'])['OrderQuantity'].sum()
cat_subcat_qty
```

```
Out[65]:
```

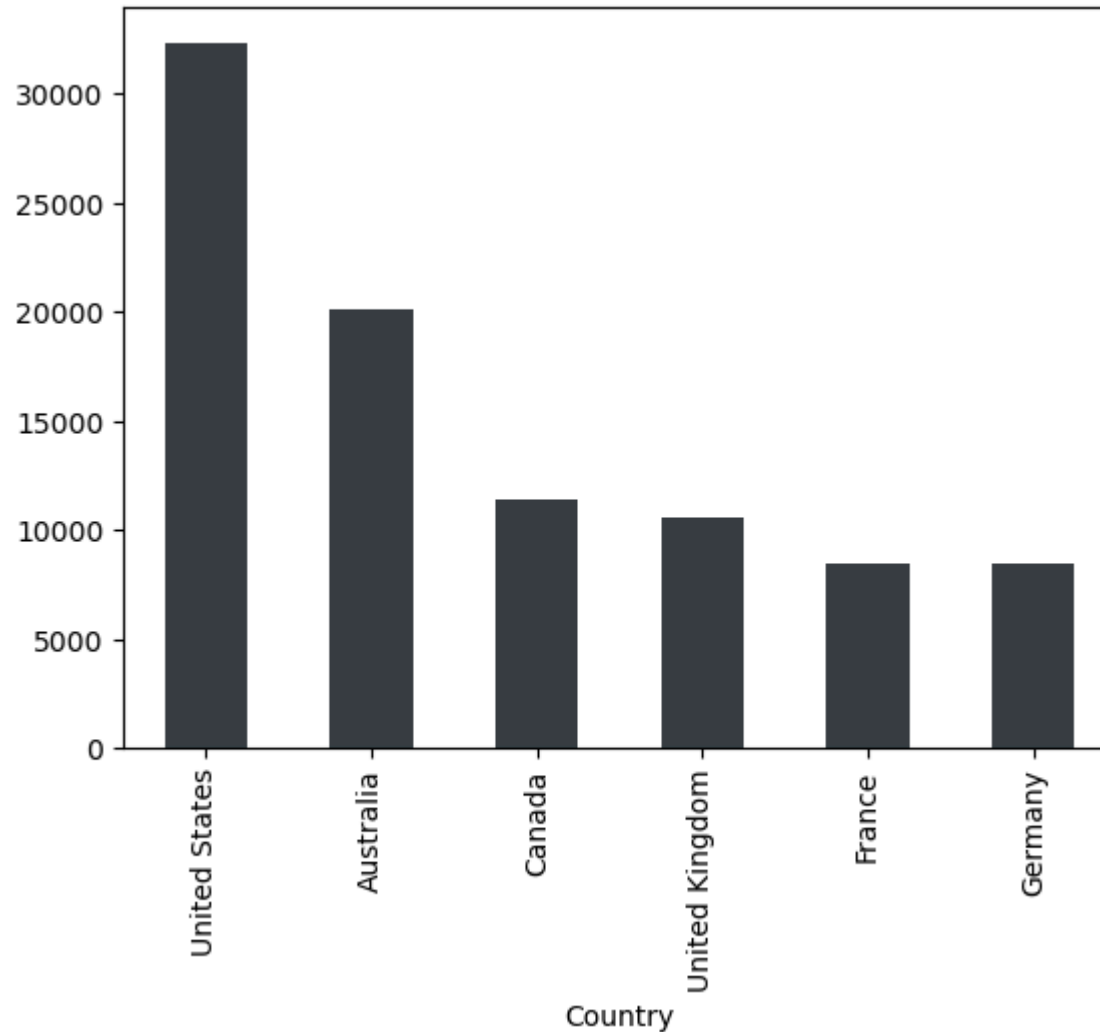
sale_year	Category	SubCategory	
2014	Bikes	Mountain Bikes	616
		Road Bikes	2876
2015	Bikes	Mountain Bikes	1661
		Road Bikes	3284
2016	Accessories	Bike Racks	493
		Bike Stands	394
		Bottles and Cages	12055
		Cleaners	1381
		Fenders	3239
		Helmets	9685
		Hydration Packs	1124
	Bikes	Tires and Tubes	25518
		Mountain Bikes	5490
		Road Bikes	6535
	Clothing	Touring Bikes	3410
		Caps	3178
		Gloves	2143
		Jerseys	5068
Shorts		1491	
Socks		856	
Vests		824	

Name: OrderQuantity, dtype: int64

Country wise quantity ordered

```
In [66]: country_qty_sales = df2.groupby('Country')['OrderQuantity'].sum().sort_values(ascending=False)
country_qty_sales.plot(kind='bar', color='#374045')
```

```
Out[66]: <Axes: xlabel='Country'>
```

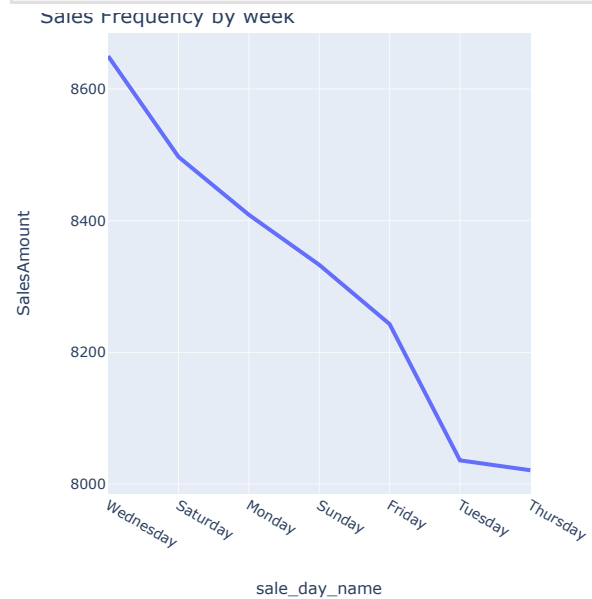


Highest quantity of products is ordered from United States.

```
In [67]: sales_by_week = df2.groupby(['sale_day_name']).count()['SalesAmount'].reset_index().sort_values('SalesAmount', ascending=False)

fig = px.line(sales_by_week, x='sale_day_name', y='SalesAmount', title='Sales Frequency by week')
fig.update_layout(
    autosize=True,
    width=300,
```

```
height=300,  
margin=dict(  
    l=25,  
    r=25,  
    b=10,  
    t=10,  
),  
font=dict(size=7))  
fig.show()
```



```
In [ ]: #High sales orders are seen on Wednesday and Saturday, therefore we can promote our product during these workweek
```

```
In [68]: df2
```

Out[68]:

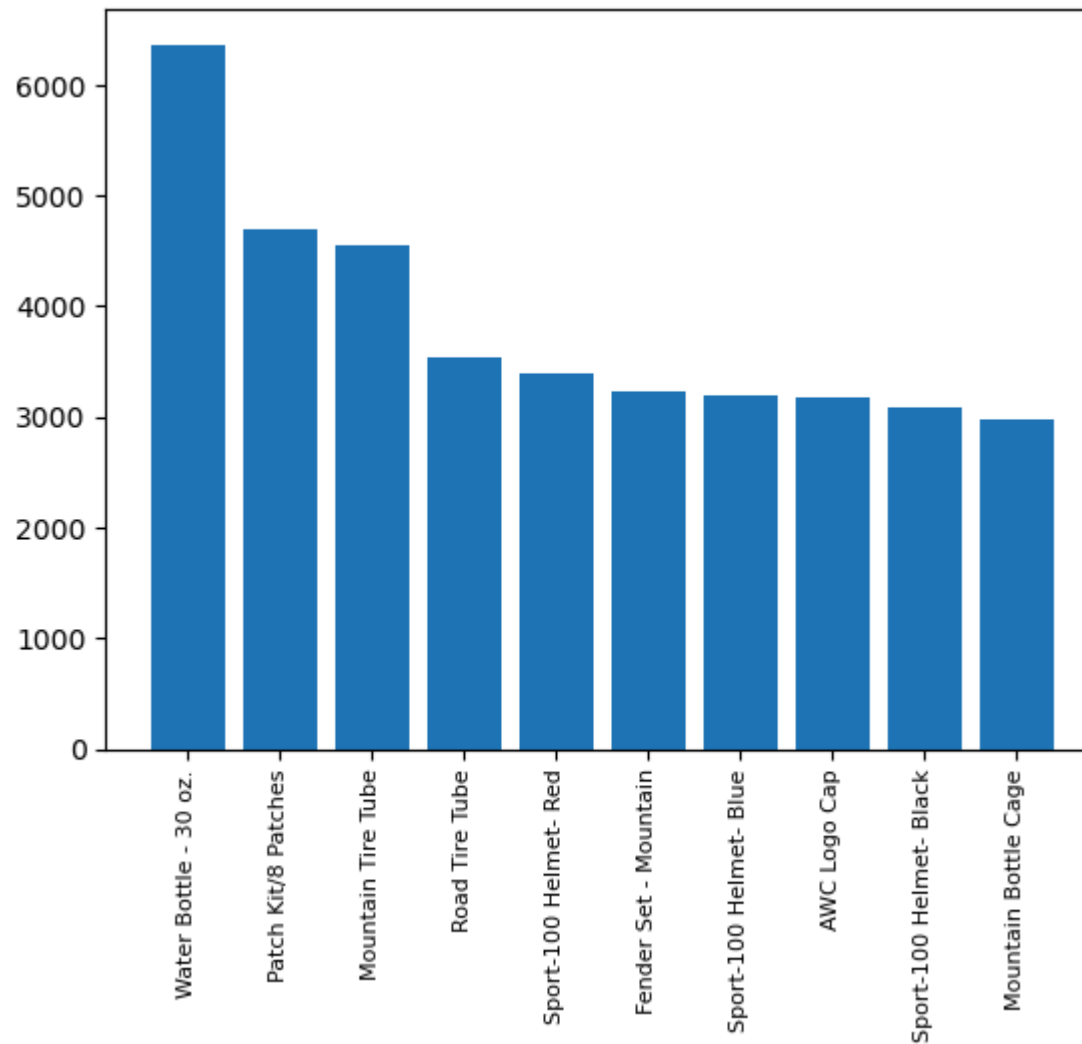
	ProductKey	OrderDate	ShipDate	CustomerKey	PromotionKey	SalesTerritoryKey	SalesOrderNumber	SalesOrderLineNumber	OrderQuantity	Un
0	310	2014-01-01	2014-01-08	21768	1	6	SO43697	1	2	178
1	600	2016-04-16	2016-04-23	21768	1	6	SO56212	1	1	53
2	310	2014-01-30	2014-02-06	21727	1	6	SO43833	1	4	89
3	479	2016-11-29	2016-12-05	21727	1	6	SO71614	2	1	
4	477	2016-11-29	2016-12-05	21727	1	6	SO71614	3	1	
...
58184	528	2016-11-07	2016-11-14	13145	1	2	SO70064	2	1	
58185	361	2016-11-07	2016-11-14	13145	1	2	SO70064	1	1	229
58186	480	2016-11-07	2016-11-14	13145	1	2	SO70064	4	1	
58187	530	2016-02-06	2016-02-13	27040	1	2	SO52124	1	1	
58188	480	2016-02-06	2016-02-13	27040	2	2	SO52124	2	1	

58189 rows × 58 columns

Which product sold the most? why do you think it sold the most?

```
In [69]: product_group = df2.groupby('ProductName')
quantity_ordered = product_group['OrderQuantity'].sum().sort_values(ascending=False)[:10]
products = quantity_ordered.index.tolist()
```

```
plt.bar(products, quantity_ordered )  
plt.xticks(products, rotation='vertical', size=8)  
plt.show()
```



Compare most ordered product by gender

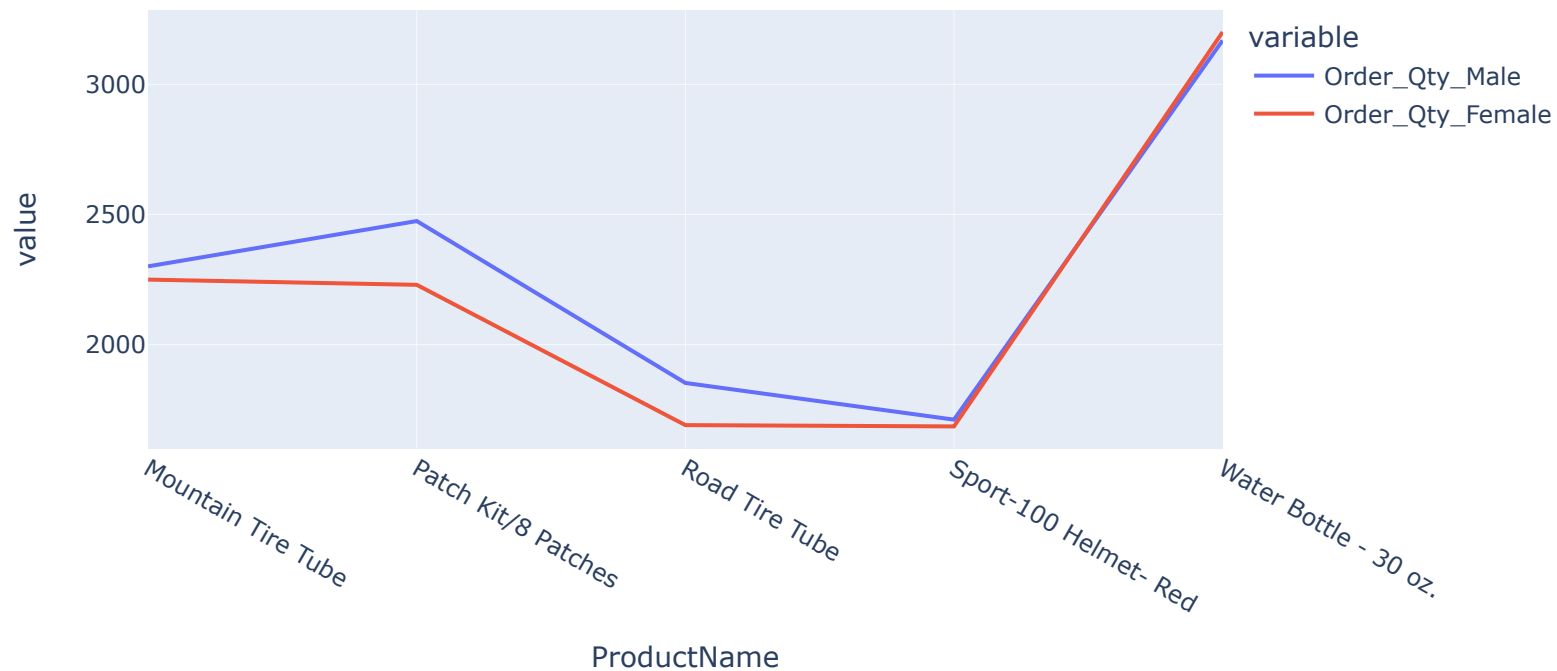
```
In [70]: male = df2[df2["Gender"]=="M"]  
female = df2[df2["Gender"]=="F"]
```

```
In [71]: male_ord_qty = male.groupby(['ProductName'],as_index=False)['OrderQuantity'].sum().nlargest(5,'OrderQuantity').sort_values('Product
male_ord_qty.columns=['ProductName', 'Order_Qty_Male']

female_ord_qty = female.groupby(['ProductName'],as_index=False)['OrderQuantity'].sum().nlargest(5,'OrderQuantity').sort_values('P
female_ord_qty.columns=['ProductName', 'Order_Qty_Female']

df2_merge = pd.merge(male_ord_qty, female_ord_qty, on='ProductName')

In [72]: fig = px.line(df2_merge, x="ProductName", y=["Order_Qty_Male","Order_Qty_Female"])
fig.update_layout(
    autosize=True,
    width=800,
    height=400)
fig.show()
```



```
In [73]: # calculating recency for customers who had made a purchase with a company
```

```
df_recency = df2.groupby(by='FullName',
                        as_index=False)['OrderDate'].max()
df_recency.columns = ['CustomerName', 'LastPurchaseDate']
recent_date = df_recency['LastPurchaseDate'].max()
df_recency['Recency'] = df_recency['LastPurchaseDate'].apply(
    lambda x: (recent_date - x).days)
```

In [74]: `df_recency['Recency'].head()`

Out[74]:

0	244
1	15
2	126
3	911
4	83

Name: Recency, dtype: int64

In [75]: *# Recency means How recently has the customer made a transaction with us*

In [76]: *# calculating the frequency of frequent transactions of the
customer in ordering/buying some product from the company.*

```
frequency_df = df2.drop_duplicates().groupby(
    by=['FullName'], as_index=False)['OrderDate'].count()
frequency_df.columns = ['CustomerName', 'Frequency']
frequency_df.head()
```

Out[76]:

	CustomerName	Frequency
0	Adams, Aaron	4
1	Adams, Adam	2
2	Adams, Alex	2
3	Adams, Alexandra	1
4	Adams, Allison	3

In [77]: *# Monetary: How much does the customer spend on purchasing products from us*

```
monetary_df = df2.groupby('FullName', as_index=False)['SalesAmount'].sum()
monetary_df.columns = ['CustomerName', 'Monetary']
monetary_df.head()
```

Out[77]:

	CustomerName	Monetary
0	Adams, Aaron	117.96
1	Adams, Adam	141.98
2	Adams, Alex	1735.98
3	Adams, Alexandra	3578.27
4	Adams, Allison	1602.47

In [78]:

df2.dtypes


```
Out[78]: ProductKey          int64
OrderDate      datetime64[ns]
ShipDate       datetime64[ns]
CustomerKey    int64
PromotionKey   int64
SalesTerritoryKey int64
SalesOrderNumber object
SalesOrderLineNumber int64
OrderQuantity  int64
UnitPrice      float64
TotalProductCost float64
SalesAmount    float64
TaxAmt         float64
StandardCost_x float64
List Price     float64
diif std cost  int64
diff list price int64
ProductName    object
SubCategory    object
Category       object
StandardCost_y float64
ListPrice      float64
DaysToManufacture int64
ProductLine    object
ModelName      object
Photo          object
ProductDescription object
StartDate      datetime64[ns]
FirstName      object
LastName       object
FullName       object
BirthDate      datetime64[ns]
MaritalStatus  object
Gender         object
YearlyIncome   int64
TotalChildren  int64
NumberChildrenAtHome int64
Education      object
Occupation     object
HouseOwnerFlag int64
NumberCarsOwned int64
AddressLine1   object
DateFirstPurchase datetime64[ns]
CommuteDistance object
```

```
Region          object
Country         object
Group           object
RegionImage     object
sale_year       int32
sale_month      int32
sale_day        int32
sale_week       int32
sale_day_name   object
year_month      object
total_Invoice_amount float64
profit          float64
Age             int32
agerange        category
dtype: object
```

```
In [ ]: # Count the customer Id
```

```
In [79]: len(df2['CustomerKey'].unique())
```

```
Out[79]: 17918
```

```
In [ ]: # Total 17918 Customer Id's are there.
```

```
In [ ]:
```

```
In [80]: df2.groupby(['CustomerKey', 'FullName'])['OrderQuantity'].sum().sort_values(ascending=False)
```

```
Out[80]: CustomerKey  FullName      OrderQuantity
11200      Griffin, Jason      115
11300      Barnes, Fernando    112
11331      Jenkins, Samantha   107
11262      Simmons, Jennifer   100
11277      Jackson, Charles    95
...
27286      Pal, Shawn          1
27289      Long, Anna          1
27290      Evans, Marcus       1
18964      Gonzalez, Thomas    1
29483      Navarro, Jesús      1
Name: OrderQuantity, Length: 17918, dtype: int64
```

```
In [81]: # Griffin Jason is the person who ordered highest quantity .
```

```
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