

▼ Import Libraries to perform specific task such as data manipulation, data analysis, numerical computations, creating various types of plots, statistical visualizations of plots.

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

▼ Load Data (Transaction) and save in variable called "tran".

```
tran = pd.read_csv("/content/transaction.csv")
tran
```

| | transaction_id | product_id | customer_id | transaction_date | online_order | ord |
|-------|----------------|------------|-------------|------------------|--------------|-----|
| 0 | 1 | 2 | 2950 | 25/02/2017 | False | |
| 1 | 2 | 3 | 3120 | 21/05/2017 | True | |
| 2 | 3 | 37 | 402 | 16/10/2017 | False | |
| 3 | 4 | 88 | 3135 | 31/08/2017 | False | |
| 4 | 5 | 78 | 787 | 01/10/2017 | True | |
| ... | ... | ... | ... | ... | ... | ... |
| 19995 | 19996 | 51 | 1018 | 24/06/2017 | True | |

▼ Some Basic Operations.

Following command gives whole information of given dataset.

```
tran.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20000 entries, 0 to 19999
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
---  -
0   transaction_id         20000 non-null  int64
1   product_id            20000 non-null  int64
2   customer_id           20000 non-null  int64
3   transaction_date       20000 non-null  object
4   online_order          19640 non-null  object
5   order_status          20000 non-null  object
6   brand                 19803 non-null  object
7   product_line          19803 non-null  object
8   product_class         19803 non-null  object
9   product_size          19803 non-null  object
10  list_price            20000 non-null  float64
11  standard_cost         19803 non-null  object
12  product_first_sold_date 19803 non-null  float64
dtypes: float64(2), int64(3), object(8)
memory usage: 2.0+ MB
```

Following command calculates order pair of given dataset.

```
tran.shape

(20000, 13)
```

Following command tells if there is any missing value in given dataset.

```
tran.isnull().sum()

transaction_id      0
product_id          0
customer_id         0
transaction_date    0
online_order       360
order_status        0
brand              197
product_line        197
product_class       197
product_size        197
list_price          0
standard_cost       197
product_first_sold_date 197
dtype: int64
```

Following command drops missing values of given dataset.

```
tran.dropna(inplace = True)
tran.isnull().sum()

transaction_id      0
product_id          0
customer_id         0
transaction_date    0
online_order        0
order_status        0
brand               0
product_line        0
product_class       0
product_size        0
list_price          0
standard_cost       0
product_first_sold_date 0
dtype: int64
```

Following command calculate order pair of given dataset.
After dropping missing values.

```
tran.shape

(19445, 13)
```

Following command gives column names of entire dataset.

```
tran.columns

Index(['transaction_id', 'product_id', 'customer_id', 'transaction_date',
      'online_order', 'order_status', 'brand', 'product_line',
      'product_class', 'product_size', 'list_price', 'standard_cost',
      'product_first_sold_date'],
      dtype='object')
```

Following command Explore "online_order" column.

```
tran.online_order.value_counts()

True      9739
False     9706
Name: online_order, dtype: int64
```

Following command Explore "order_status" column.

```
tran.order_status.value_counts()

Approved      19273
Cancelled       172
Name: order_status, dtype: int64
```

Following command Explore "product_class" column.

```
tran.product_class.value_counts()

medium      13587
high        2952
low         2906
Name: product_class, dtype: int64
```

Following command Explore "product_size" column.

```
tran.product_size.value_counts()

medium      12767
large        3900
small        2778
Name: product_size, dtype: int64
```

Following command calculate duplicate value in given dataset, if exists.

```
tran.duplicated().sum()

0
```

▼ Convert Float to datetime at first_sold column

Following command gives data type of "first_sold" column before converting in given dataset.

```
tran.product_first_sold_date.dtype

dtype('float64')
```

Following command replace data type of first_sold column in given dataset.

```
tran['product_first_sold_date'] = pd.to_datetime(tran['product_first_sold_date'], unit='s')
```

Following command return "first_sold" column after replacing.

```
tran.product_first_sold_date

0      1970-01-01 11:27:25
1      1970-01-01 11:35:01
2      1970-01-01 10:06:01
3      1970-01-01 10:02:25
4      1970-01-01 11:43:46
...
19995   1970-01-01 10:30:23
19996   1970-01-01 09:52:40
19997   1970-01-01 11:13:30
19998   1970-01-01 10:36:56
19999   1970-01-01 10:05:34
Name: product_first_sold_date, Length: 19445, dtype: datetime64[ns]
```

Following command gives data type of "first_sold" column after converting in given dataset.

```
tran.product_first_sold_date.dtype

dtype('<M8[ns]')
```

▼ Add currency in price_list column of Given dataset

Following command return "list_price" column before editing.

```
tran.list_price

0      71.49
1     2091.47
2     1793.43
3     1198.46
4     1765.30
...
19995   2005.66
19996    416.98
19997   1636.90
19998    227.88
19999   1775.81
Name: list_price, Length: 19445, dtype: float64
```

Following command edit "list_price" column, add \$ sign

```
tran["list_price"] = tran["list_price"].map("${:,.0f}").format)
```

Following command return "list_price" column after editing.

```
tran.list_price

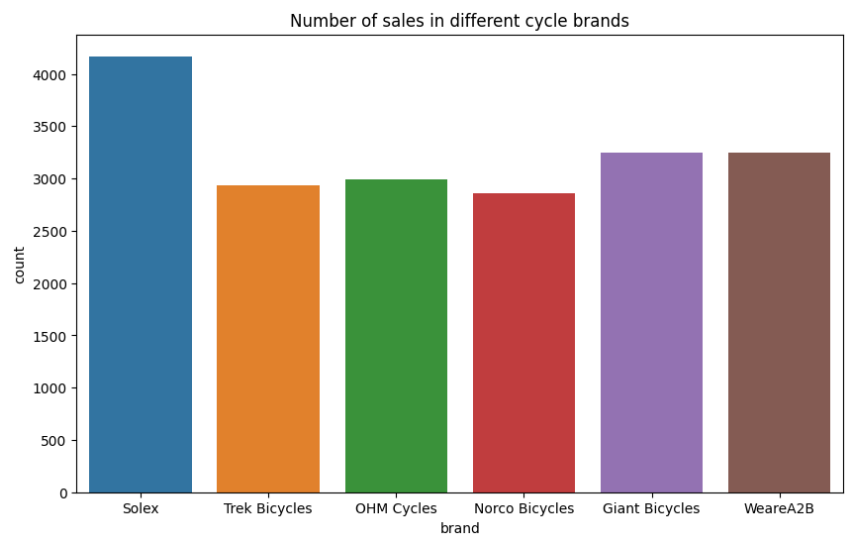
0      $71
1     $2,091
2     $1,793
3     $1,198
4     $1,765
```

```
...
19995    $2,006
19996     $417
19997    $1,637
19998     $228
19999    $1,776
Name: list_price, Length: 19445, dtype: object
```

▼ Graphical view

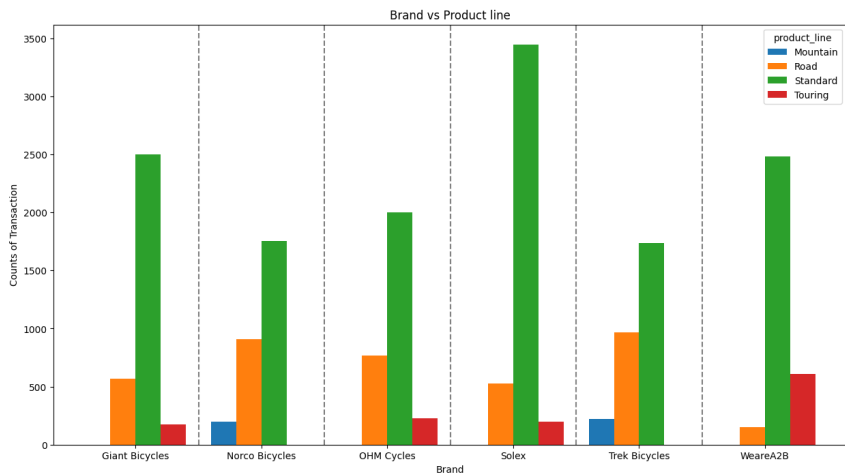
Following command shows the graphical view of "Brand" column

```
plt.figure(figsize=(10,6))
count = sns.countplot(x= tran['brand'], data=tran)
plt.title("Number of sales in different cycle brands")
plt.show()
```



Following commands shows the relationship between different brands with product line

```
counts = tran.groupby(['brand', 'product_line']).size().reset_index(name='Count')
pivot_counts = counts.pivot(index='brand', columns='product_line', values='Count')
# Create a bar plot
pivot_counts.plot(kind='bar', figsize=(15, 8), width=0.8)
# Creating a line
separation_lines = [1, 2, 3, 4, 5]
for line in separation_lines:
    plt.axvline(x=line - 0.5, color='grey', linestyle='--')
plt.xlabel('Brand')
plt.ylabel('Counts of Transaction')
plt.title('Brand vs Product line')
plt.xticks(rotation = 0)
plt.show()
```



Following commands shows the relationship between different brands with product class

```
counts = tran.groupby(['brand', 'product_class']).size().reset_index(name='Count')
# Pivot the data for plotting
pivot_counts = counts.pivot(index='brand', columns='product_class', values='Count')
# Create a bar plot
pivot_counts.plot(kind='bar', figsize=(15, 8), width=0.8)
# Creating a line
separation_lines = [1, 2, 3, 4, 5]
for line in separation_lines:
    plt.axvline(x=line - 0.5, color='grey', linestyle='--')
# Set labels and title
plt.xlabel('Brand')
plt.ylabel('Counts of Transaction')
plt.title('Brand vs Product Class')
plt.xticks(rotation = 0)
plt.show()
```

