

```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
```

```
In [6]: df = pd.read_csv('Amazon Sale Report.csv', encoding= 'unicode escape')
```

```
In [7]: df.shape
```

Out[7]: (128976, 21)

```
In [8]: df.head()
```

Out[8]:

	index	Order ID	Date	Status	Fulfilment	Sales Channel	ship-service-level	Category	Size	Courier Status	...	currency	Amount	ship-ci
0	0	405-8078784-5731545	04-30-22	Cancelled	Merchant	Amazon.in	Standard	T-shirt	S	On the Way	...	INR	647.62	MUMB
1	1	171-9198151-1101146	04-30-22	Shipped - Delivered to Buyer	Merchant	Amazon.in	Standard	Shirt	3XL	Shipped	...	INR	406.00	BENGALUF
2	2	404-0687676-7273146	04-30-22	Shipped	Amazon	Amazon.in	Expedited	Shirt	XL	Shipped	...	INR	329.00	NAVI MUMB
3	3	403-9615377-8133951	04-30-22	Cancelled	Merchant	Amazon.in	Standard	Blazzer	L	On the Way	...	INR	753.33	PUDUCHERF
4	4	407-1069790-7240320	04-30-22	Shipped	Amazon	Amazon.in	Expedited	Trousers	3XL	Shipped	...	INR	574.00	CHENN

5 rows × 21 columns

```
In [9]: df.tail()
```

Out[9]:

	index	Order ID	Date	Status	Fulfilment	Sales Channel	ship-service-level	Category	Size	Courier Status	...	currency	Amount	shi
128971	128970	406-6001380-7673107	05-31-22	Shipped	Amazon	Amazon.in	Expedited	Shirt	XL	Shipped	...	INR	517.0	HYDER
128972	128971	402-9551604-7544318	05-31-22	Shipped	Amazon	Amazon.in	Expedited	T-shirt	M	Shipped	...	INR	999.0	GURU
128973	128972	407-9547469-3152358	05-31-22	Shipped	Amazon	Amazon.in	Expedited	Blazzer	XXL	Shipped	...	INR	690.0	HYDER
128974	128973	402-6184140-0545956	05-31-22	Shipped	Amazon	Amazon.in	Expedited	T-shirt	XS	Shipped	...	INR	1199.0	
128975	128974	408-7436540-8728312	05-31-22	Shipped	Amazon	Amazon.in	Expedited	T-shirt	S	Shipped	...	INR	696.0	

5 rows × 21 columns

```
In [10]: df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 128976 entries, 0 to 128975
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   index                 128976 non-null  int64
1   Order ID              128976 non-null  object
2   Date                  128976 non-null  object
3   Status                128976 non-null  object
4   Fulfilment            128976 non-null  object
5   Sales Channel         128976 non-null  object
6   ship-service-level    128976 non-null  object
7   Category              128976 non-null  object
8   Size                  128976 non-null  object
9   Courier Status        128976 non-null  object
10  Qty                   128976 non-null  int64
11  currency              121176 non-null  object
12  Amount               121176 non-null  float64
13  ship-city             128941 non-null  object
14  ship-state            128941 non-null  object
15  ship-postal-code      128941 non-null  float64
16  ship-country          128941 non-null  object
17  B2B                   128976 non-null  bool
18  fulfilled-by          39263 non-null   object
19  New                   0 non-null       float64
20  PendingS              0 non-null       float64
dtypes: bool(1), float64(4), int64(2), object(14)
memory usage: 19.8+ MB

```

```

In [12]: #Removing blank columns
df.drop(['New','PendingS'], axis = 1, inplace = True)

```

```

In [13]: df.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 128976 entries, 0 to 128975
Data columns (total 19 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   index                 128976 non-null  int64
1   Order ID              128976 non-null  object
2   Date                  128976 non-null  object
3   Status                128976 non-null  object
4   Fulfilment            128976 non-null  object
5   Sales Channel         128976 non-null  object
6   ship-service-level    128976 non-null  object
7   Category              128976 non-null  object
8   Size                  128976 non-null  object
9   Courier Status        128976 non-null  object
10  Qty                   128976 non-null  int64
11  currency              121176 non-null  object
12  Amount               121176 non-null  float64
13  ship-city             128941 non-null  object
14  ship-state            128941 non-null  object
15  ship-postal-code      128941 non-null  float64
16  ship-country          128941 non-null  object
17  B2B                   128976 non-null  bool
18  fulfilled-by          39263 non-null  object
dtypes: bool(1), float64(2), int64(2), object(14)
memory usage: 17.8+ MB

```

```

In [15]: #checking null values
pd.isnull(df).sum()

```

```

Out[15]: index                0
Order ID                    0
Date                        0
Status                      0
Fulfilment                  0
Sales Channel                0
ship-service-level          0
Category                    0
Size                        0
Courier Status              0
Qty                          0
currency                    7800
Amount                      7800
ship-city                   35
ship-state                  35
ship-postal-code            35
ship-country                35
B2B                         0
fulfilled-by                89713
dtype: int64

```

In [17]:

df.shape

Out[17]:

(128976, 19)

In [18]:

#dropping null values
df.dropna(inplace = True)

In [19]:

df.shape

Out[19]:

(37514, 19)

In [21]:

df.columns

Out[21]:

Index(['index', 'Order ID', 'Date', 'Status', 'Fulfilment', 'Sales Channel',
 'ship-service-level', 'Category', 'Size', 'Courier Status', 'Qty',
 'currency', 'Amount', 'ship-city', 'ship-state', 'ship-postal-code',
 'ship-country', 'B2B', 'fulfilled-by'],
 dtype='object')

In [38]:

#changing data type
df['ship-postal-code'] = df['ship-postal-code'].astype('int')

In [39]:

df['ship-postal-code'].dtype

Out[39]:

dtype('int32')

In [40]:

df['Date'] = pd.to_datetime (df['Date'])

In [41]:

df.columns

Out[41]:

Index(['index', 'Order ID', 'Date', 'Status', 'Fulfilment', 'Sales Channel',
 'ship-service-level', 'Category', 'Size', 'Courier Status', 'Quantity',
 'currency', 'Amount', 'ship-city', 'ship-state', 'ship-postal-code',
 'ship-country', 'B2B', 'fulfilled-by'],
 dtype='object')

In [52]:

#renaming the columns
df.rename(columns = {'Qty': 'Quantity'}, inplace = True)

In [53]:

df.describe()

Out[53]:

	index	Date	Quantity	Amount	ship-postal-code
count	37514.000000	37514	37514.000000	37514.000000	37514.000000
mean	60953.809858	2022-05-11 07:56:47.303939840	0.867383	646.553960	463291.552754
min	0.000000	2022-03-31 00:00:00	0.000000	0.000000	110001.000000
25%	27235.250000	2022-04-20 00:00:00	1.000000	458.000000	370465.000000
50%	63470.500000	2022-05-09 00:00:00	1.000000	629.000000	500019.000000
75%	91790.750000	2022-06-01 00:00:00	1.000000	771.000000	600042.000000
max	128891.000000	2022-06-29 00:00:00	5.000000	5495.000000	989898.000000
std	36844.853039	NaN	0.354160	279.952414	194550.425637

In [54]:

df.describe(include = 'object')

Out[54]:

	Order ID	Status	Fulfilment	Sales Channel	ship-service-level	Category	Size	Courier Status	currency	ship-city	ship-state	ci
count	37514	37514	37514	37514	37514	37514	37514	37514	37514	37514	37514	
unique	34664	11	1	1	1	8	11	3	1	4698	58	
top	171-5057375-2831560	Shipped - Delivered to Buyer	Merchant	Amazon.in	Standard	T-shirt	M	Shipped	INR	BENGALURU	MAHARASHTRA	
freq	12	28741	37514	37514	37514	14062	6806	31859	37514	2839	6236	

In [55]:

#using describe for specific columns
df[['Quantity','Amount']].describe()

Out[55]:

	Quantity	Amount
count	37514.000000	37514.000000
mean	0.867383	646.553960
std	0.354160	279.952414
min	0.000000	0.000000
25%	1.000000	458.000000
50%	1.000000	629.000000
75%	1.000000	771.000000
max	5.000000	5495.000000

Exploratory Data Analysis

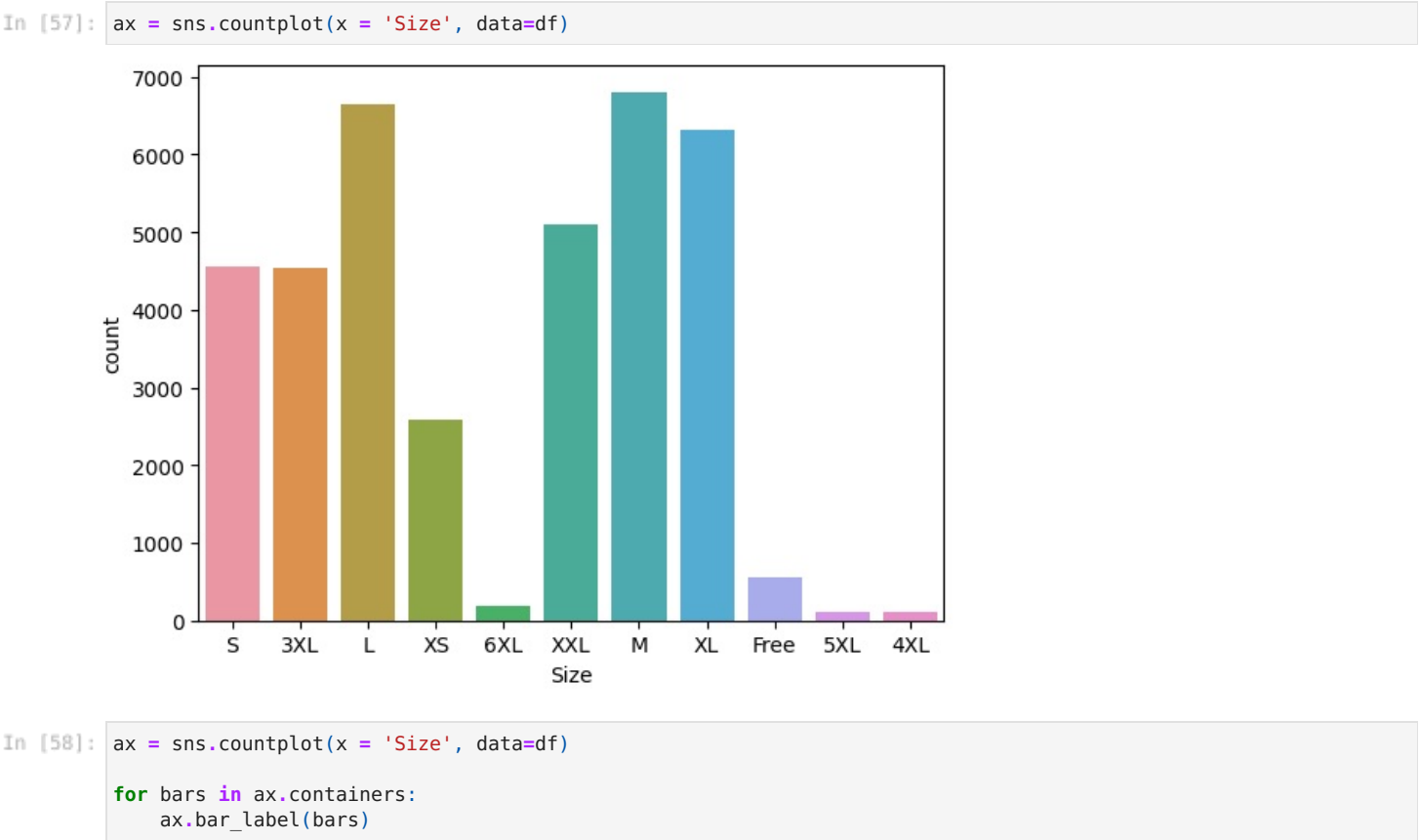
In [56]:

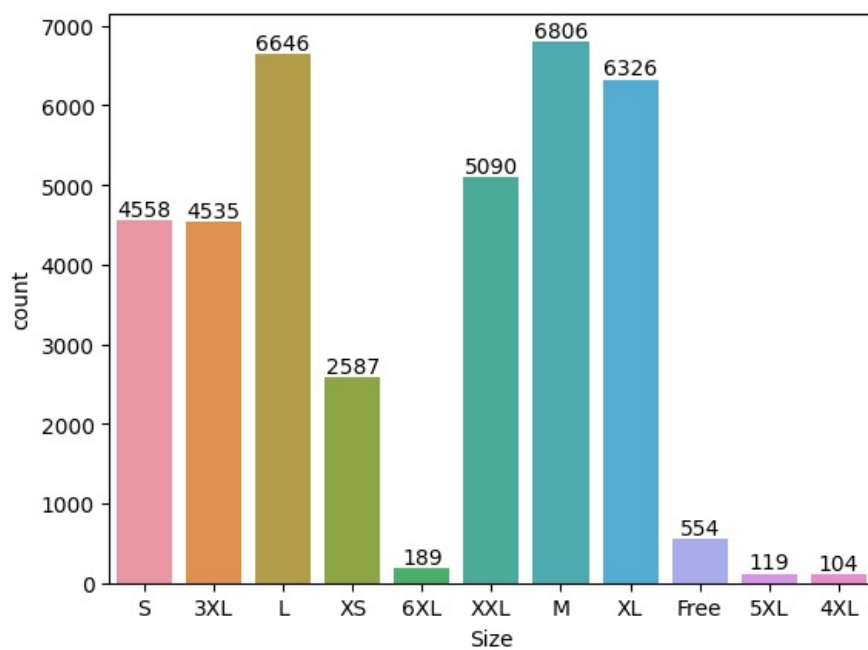
```
df.columns
```

Out[56]:

```
Index(['index', 'Order ID', 'Date', 'Status', 'Fulfilment', 'Sales Channel',  
      'ship-service-level', 'Category', 'Size', 'Courier Status', 'Quantity',  
      'currency', 'Amount', 'ship-city', 'ship-state', 'ship-postal-code',  
      'ship-country', 'B2B', 'fulfilled-by'],  
      dtype='object')
```

Size





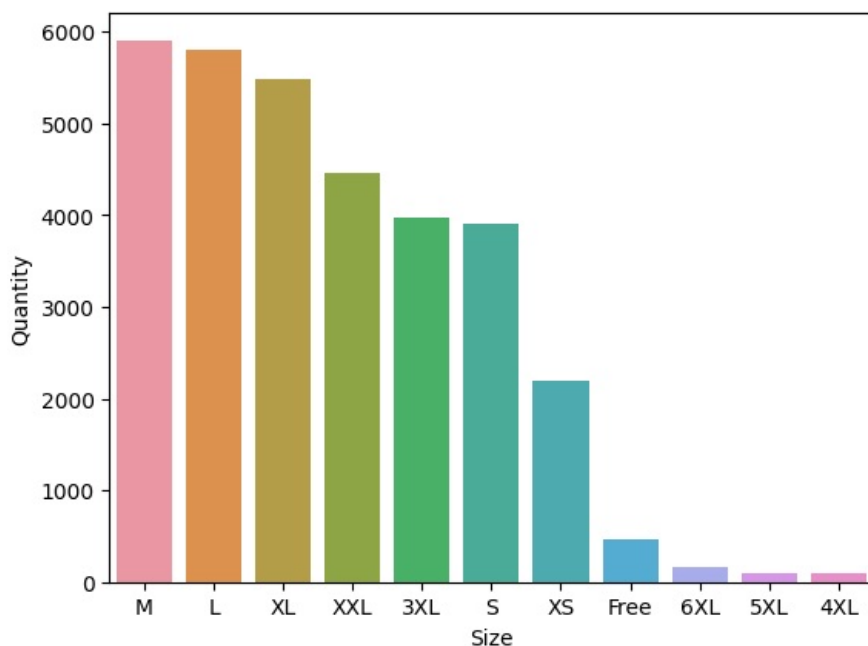
```
In [59]: df.groupby(['Size'], as_index = False)['Quantity'].sum().sort_values(by = 'Quantity',ascending = False)
```

```
Out[59]:
```

	Size	Quantity
6	M	5905
5	L	5795
8	XL	5481
10	XXL	4465
0	3XL	3972
7	S	3896
9	XS	2191
4	Free	467
3	6XL	170
2	5XL	104
1	4XL	93

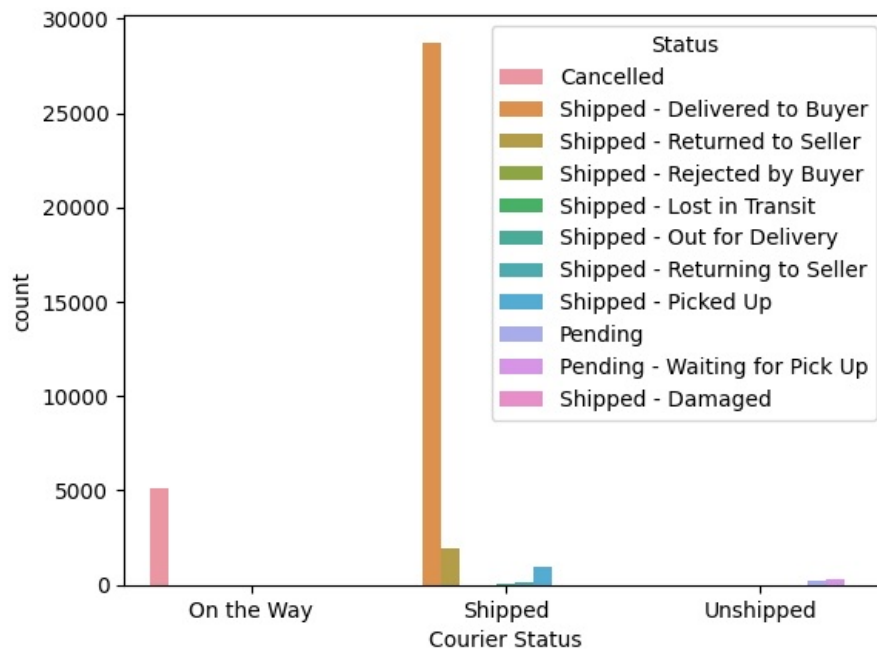
```
In [62]: S_Quantity = df.groupby(['Size'], as_index = False)['Quantity'].sum().sort_values(by = 'Quantity',ascending = False)
sns.barplot(x = 'Size', y ='Quantity', data = S_Quantity )
```

```
Out[62]: <Axes: xlabel='Size', ylabel='Quantity'>
```



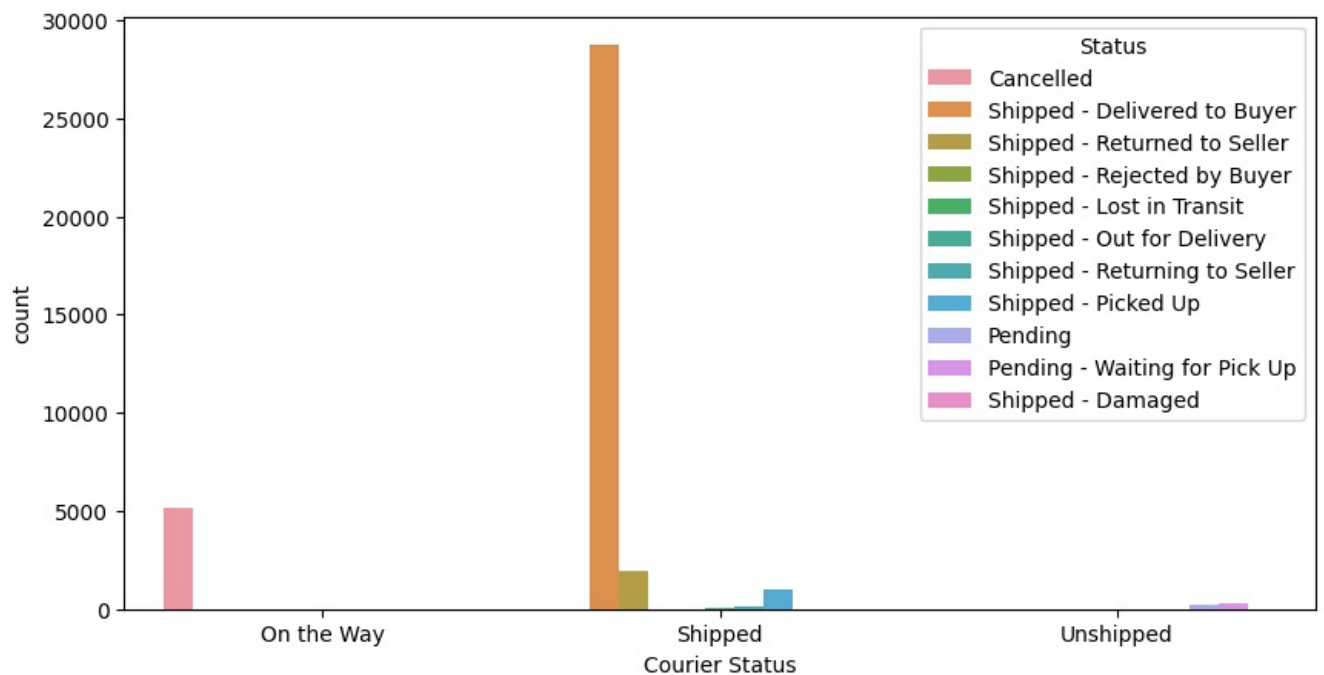
```
In [63]: sns.countplot(data = df, x = 'Courier Status', hue = 'Status')
```

```
Out[63]: <Axes: xlabel='Courier Status', ylabel='count'>
```



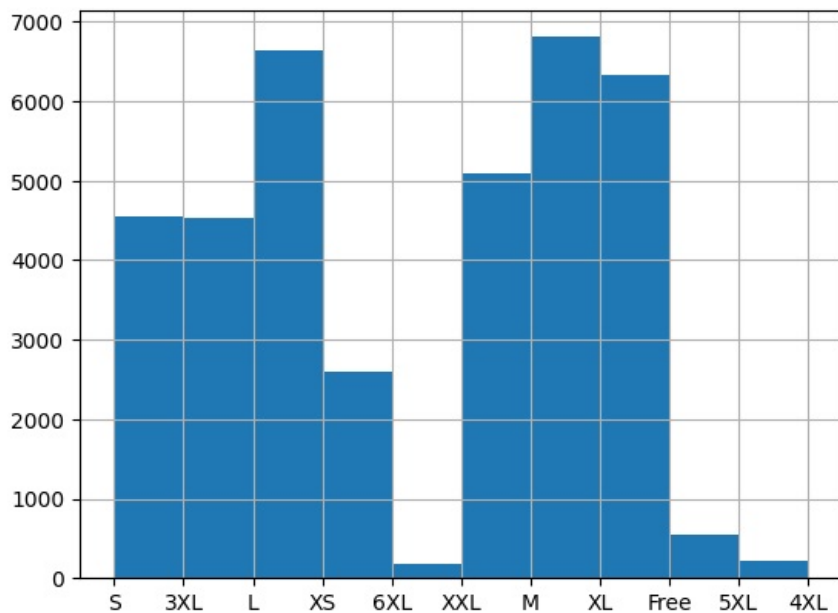
```
In [66]: plt.figure(figsize = (10,5))
```

```
ax = sns.countplot(data = df, x = 'Courier Status', hue = 'Status')  
plt.show()
```

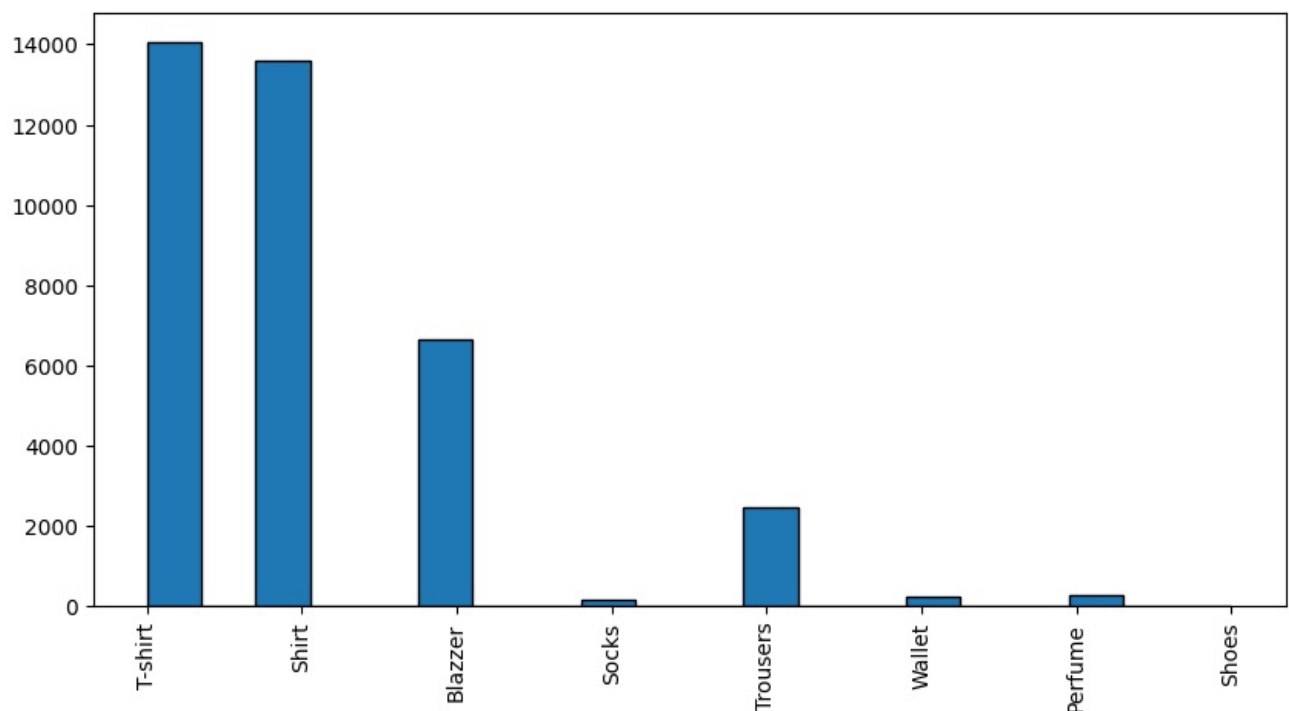


```
In [67]: df['Size'].hist()
```

```
Out[67]: <Axes: >
```

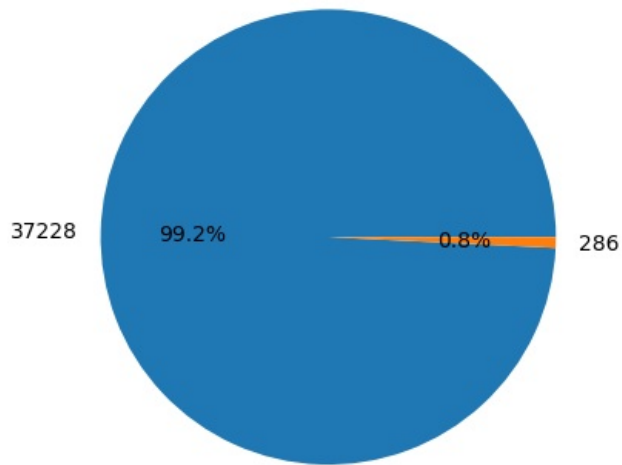


```
In [69]: df['Category'] = df['Category'].astype(str)
column_data = df['Category']
plt.figure(figsize = (10,5))
plt.hist(column_data, bins = 20, edgecolor = 'Black')
plt.xticks(rotation = 90)
plt.show()
```

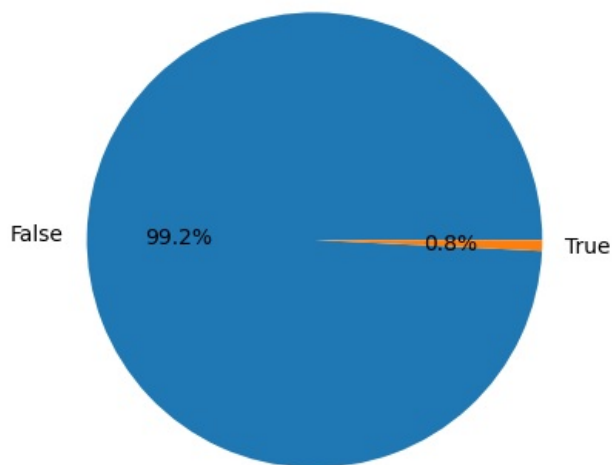


```
In [70]: #checking B2B Data by using pie chart
B2B_Check = df['B2B'].value_counts()

#plotting pie chart
plt.pie(B2B_Check, labels = B2B_Check, autopct = '%1.1f%')
plt.show()
```



```
In [71]: B2B_Check = df['B2B'].value_counts()
plt.pie(B2B_Check, labels = B2B_Check.index, autopct = '%1.1f%%')
plt.show()
```

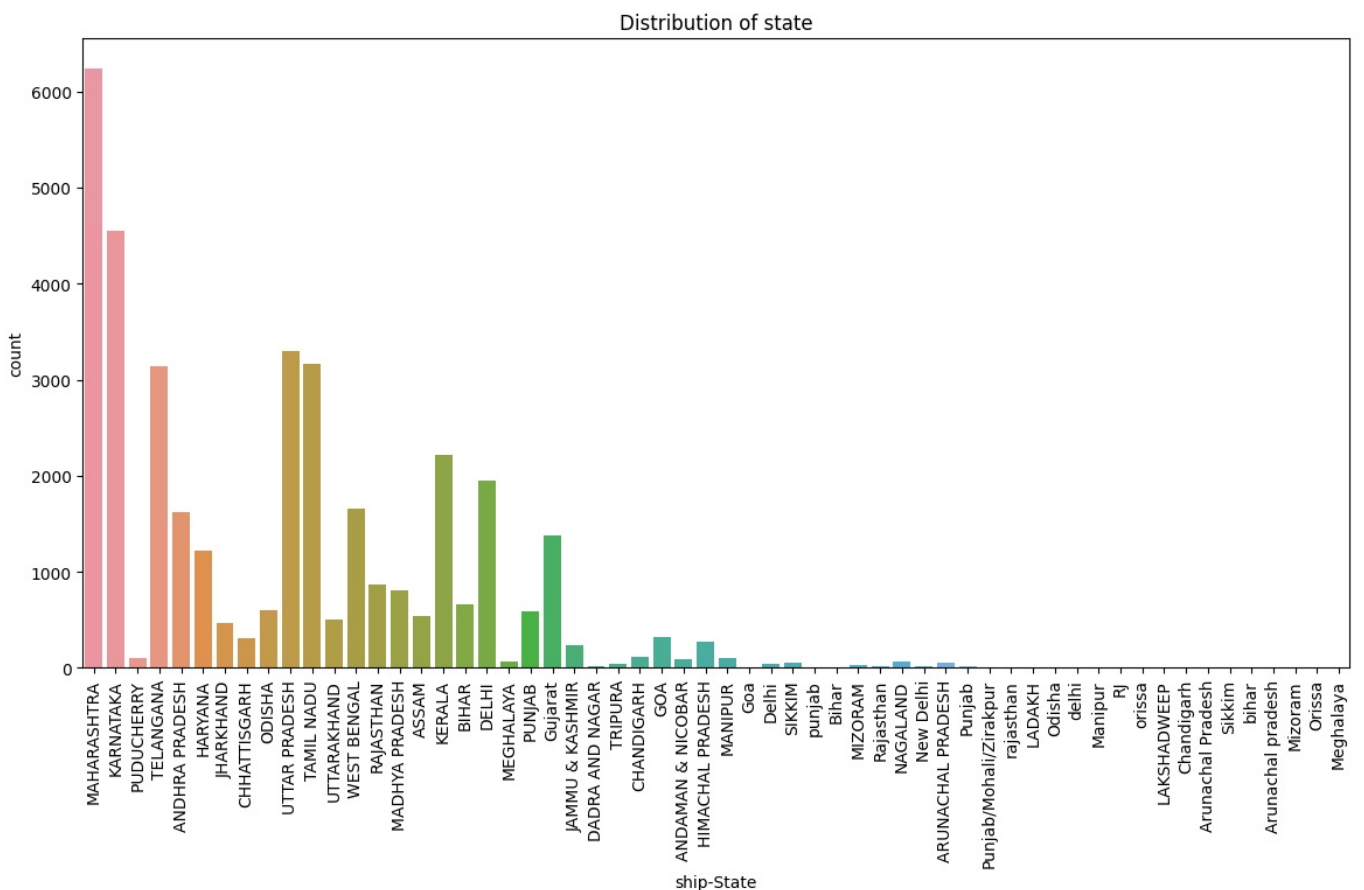


```
In [72]: #preparing data for scatter plot
x_data = df['Category']
y_data = df['Size']

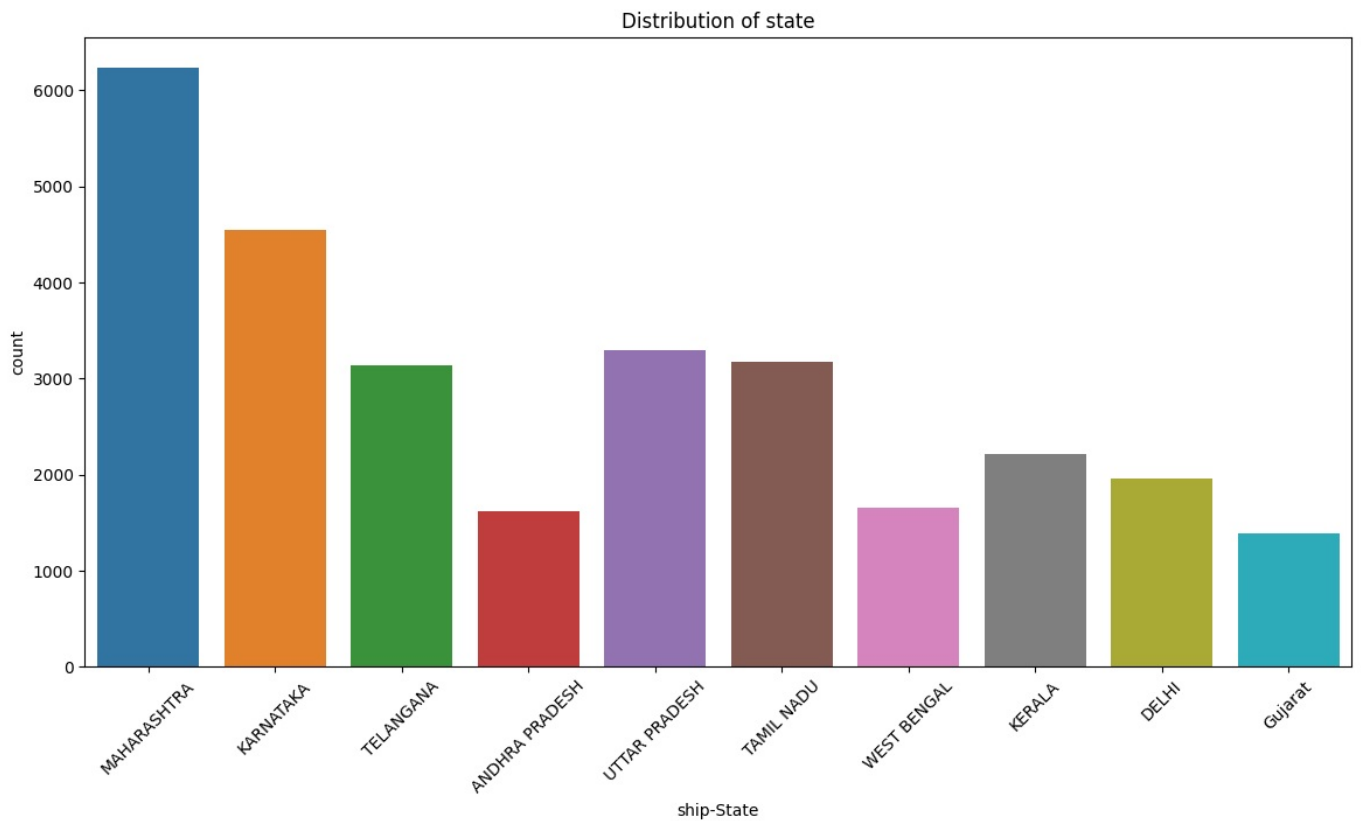
#plotting the scatter plot
plt.scatter(x_data, y_data)
plt.xlabel('Category')
plt.ylabel('Size')
plt.title('Scatter Plot')
plt.show()
```




```
In [77]: #plotting the count of cities by state
plt.figure(figsize=(14,7))
sns.countplot(data = df, x = 'ship-state')
plt.xlabel('ship-State')
plt.ylabel('count')
plt.title('Distribution of state')
plt.xticks(rotation = 90)
plt.show()
```



```
In [81]: Top_10_State = df['ship-state'].value_counts().head(10)
plt.figure(figsize=(14,7))
sns.countplot(data = df[df['ship-state'].isin(Top_10_State.index)], x = 'ship-state')
plt.xlabel('ship-State')
plt.ylabel('count')
plt.title('Distribution of state')
plt.xticks(rotation = 45)
plt.show()
```



Insights

Most of the People Buys M-size

Most of the Quantity Buys M-size in sales

Majority of the Orders are shipped through courier

Most of the People buys T-shirt

Maximum(99.2%) of the buyers are retailers and 0.8% are B2B buyers

Most of the buyers are from Maharashtra State

Conclusion

The data analysis reveals that the buisness has a significant customer base in Maharashtra state, mainly serves retailers, Fulfil orders through Amazon, experinces high demand for T-shirts and sees M-size as the preffered choice among buyers.

In []:

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js