```
In [1]:
           import pandas as pd
           import numpy as np
           import seaborn as sns
           import matplotlib.pyplot as plt
 In [3]: data = pd.read csv('Cleaned Customer.csv')
 In [7]:
           data
 Out[7]:
                  Unnamed:
                             CustomerID
                                          Age Gender ContractType MonthlyCharges TotalCharges TechSupport In-
               0
                          0
                                    1083
                                          79.0
                                                  Male
                                                             One year
                                                                            90.038513
                                                                                        3511.502019
                                                                                                              No
               1
                          1
                                    1117
                                          60.0
                                                Female
                                                             One year
                                                                            80.590894
                                                                                        2901.272196
                                                                                                              No
                          2
               2
                                     437
                                          53.0
                                                Female
                                                             Two year
                                                                            98.695968
                                                                                        6513.933908
                                                                                                              Yes
               3
                          3
                                    3833
                                          84.0
                                                Female
                                                             One year
                                                                            43.042067
                                                                                        1549.514395
                                                                                                              No
               4
                          4
                                    1976
                                          69.0
                                                  Male
                                                             One year
                                                                            51.930032
                                                                                        2232.991377
                                                                                                              No
                                                                                   ...
                                                                                                               ...
                                          54.0
            4646
                       5015
                                    1514
                                                  Male
                                                             One year
                                                                            57.803077
                                                                                         462.424613
                                                                                                              No
                                    2716 45.0
            4647
                       5016
                                                  Male
                                                             Two year
                                                                           103.314530
                                                                                         826.516243
                                                                                                              No
            4648
                                          21.0
                       5017
                                     756
                                                Female
                                                             Two year
                                                                           103.105344
                                                                                         103.105344
                                                                                                              Yes
            4649
                                    3284
                                          85.0
                       5018
                                                  Male
                                                             Two year
                                                                            36.907180
                                                                                        1660.823112
                                                                                                              Yes
            4650
                       5019
                                          65.0
                                                Female
                                                                            98.243941
                                                                                        6582.344044
                                      96
                                                             One year
                                                                                                              No
           4651 rows × 13 columns
           data.drop(columns=['Unnamed: 0'],inplace=True)
 In [9]:
In [11]:
           data.sample(5)
Out[11]:
                               Age
                  CustomerID
                                    Gender
                                            ContractType MonthlyCharges
                                                                            TotalCharges TechSupport InternetService
            3823
                          769
                               43.0
                                       Male
                                                  Two year
                                                                 71.425428
                                                                             3285.569684
                                                                                                  Yes
                                                                                                            Fiber optic
            1698
                               85.0
                                                                 85.809256
                          236
                                       Male
                                                  Two year
                                                                             1630.375866
                                                                                                   No
                                                                                                                   No
                                                 Month-to-
            1116
                         2159
                               44.0
                                       Male
                                                                106.302596
                                                                             4252.103844
                                                                                                   Yes
                                                                                                                 DSI
                                                    month
             250
                          997
                               51.0
                                     Female
                                                  Two year
                                                                108.922689
                                                                             2940.912607
                                                                                                   Yes
                                                                                                                   No
            2864
                         1666 58.0
                                       Male
                                                 One year
                                                                100.211645
                                                                              400.846579
                                                                                                   Yes
                                                                                                            Fiber optic
           df = data.copy()
In [13]:
```

```
In [18]: df.info()
```

```
RangeIndex: 4651 entries, 0 to 4650
Data columns (total 12 columns):
     Column
                       Non-Null Count
                                       Dtype
_ _ _
0
     CustomerID
                       4651 non-null
                                       int64
1
     Age
                       4651 non-null
                                       float64
                                       object
2
     Gender
                       4651 non-null
3
     ContractType
                       4651 non-null
                                       object
                                       float64
4
    MonthlyCharges
                       4651 non-null
5
     TotalCharges
                       4651 non-null
                                       float64
6
    TechSupport
                       4651 non-null
                                       object
7
     InternetService
                                       object
                       4651 non-null
8
     Tenure
                                       int64
                       4651 non-null
9
     PaperlessBilling 4651 non-null
                                       object
10 PaymentMethod
                       4651 non-null
                                       object
11 Churn
                       4651 non-null
                                       object
dtypes: float64(3), int64(2), object(7)
memory usage: 436.2+ KB
```

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

Column Types

- Numerical CustomerID, Age, Monthlychargers, Totalcharges, Tenure
- Categorical Gender, ContractType, TechSupport, InternetService, PaperlessBilling,PaymentMethod,Churn

Univariate Analysis

Univariate analysis focuses on analyzing each feature in the dataset independently.

- **Distribution analysis**: The distribution of each feature is examined to identify its shape, central tendency, and dispersion.
- **Identifying potential issues**: Univariate analysis helps in identifying potential problems with the data such as outliers, skewness, and missing values

Univariate Analysis on Categorical Columns

```
In [14]: df.info()
         <class 'pandas.core.frame.DataFrame'>
```

RangeIndex: 4651 entries, 0 to 4650 Data columns (total 12 columns): # Column Non-Null Count Dtype ---0 CustomerID int64 4651 non-null 1 Age 4651 non-null float64 2 Gender 4651 non-null object object 3 ContractType 4651 non-null MonthlyCharges 4651 non-null float64 5 TotalCharges float64 4651 non-null 6 TechSupport 4651 non-null object 7 InternetService 4651 non-null object int64 8 Tenure 4651 non-null 9 PaperlessBilling 4651 non-null object 10 PaymentMethod 4651 non-null object 11 Churn 4651 non-null object dtypes: float64(3), int64(2), object(7)

memory usage: 436.2+ KB

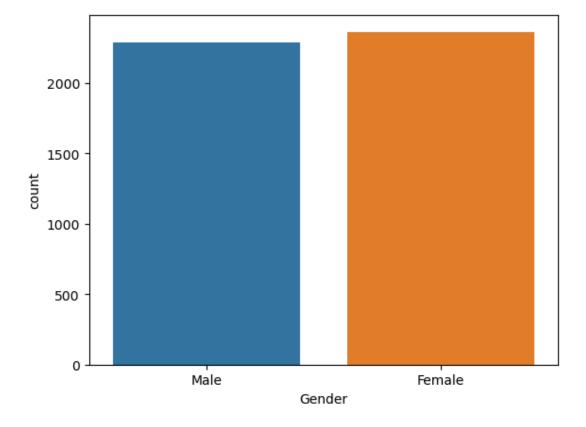
Gender

Conclusion :-

- No Missing Values
- Female Customer are More in number

```
In [19]: sns.countplot(data=df,x='Gender')
```

Out[19]: <Axes: xlabel='Gender', ylabel='count'>



```
In [20]: df.Gender.isna().sum()
```

Out[20]: 0

ContractType:

Conclusion:-

- No Missing Values
- Converted to new Column ContractType_in_Days which contains the number of days as a contract

```
In [21]: df.ContractType.value_counts()
```

Out[21]: ContractType

One year 1595
Two year 1556
Month-to-month 1500
Name: count, dtype: int64

```
In [26]:
         def conversion days(data):
             if 'One year' == data:
                 return 365
             elif 'Two year' == data:
                 return 365*2
             else:
                 return 30
         days = df.ContractType.apply(conversion_days)
In [28]: |df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 4651 entries, 0 to 4650
         Data columns (total 12 columns):
              Column
                                Non-Null Count
                                                Dtype
         ---
              _____
                                -----
                                                ----
          0
              CustomerID
                                                int64
                                4651 non-null
          1
              Age
                                4651 non-null
                                                float64
          2
              Gender
                                4651 non-null
                                                object
              ContractType
                                4651 non-null
                                                object
          4
                                                float64
              MonthlyCharges
                                4651 non-null
          5
              TotalCharges
                                4651 non-null
                                                float64
          6
              TechSupport
                                4651 non-null
                                                object
          7
              InternetService
                                4651 non-null
                                                object
              Tenure
                                4651 non-null
                                                int64
          9
              PaperlessBilling 4651 non-null
                                                object
          10 PaymentMethod
                                4651 non-null
                                                object
          11 Churn
                                4651 non-null
                                                object
         dtypes: float64(3), int64(2), object(7)
         memory usage: 436.2+ KB
In [29]: | df.insert(3, 'ContractType In days', days)
In [31]: | df.drop(columns=['ContractType'],inplace=True)
In [32]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 4651 entries, 0 to 4650
         Data columns (total 12 columns):
          #
              Column
                                    Non-Null Count
                                                    Dtype
         ---
              _____
                                     _____
          0
              CustomerID
                                    4651 non-null
                                                    int64
                                                    float64
          1
              Age
                                    4651 non-null
          2
              Gender
                                    4651 non-null
                                                    object
          3
              ContractType In days
                                    4651 non-null
                                                    int64
                                                    float64
          4
              MonthlyCharges
                                    4651 non-null
              TotalCharges
                                    4651 non-null
                                                    float64
          6
              TechSupport
                                    4651 non-null
                                                    object
              InternetService
                                                    object
                                    4651 non-null
          8
              Tenure
                                    4651 non-null
                                                    int64
              PaperlessBilling
          9
                                    4651 non-null
                                                    object
          10 PaymentMethod
                                    4651 non-null
                                                    object
          11 Churn
                                    4651 non-null
                                                    object
         dtypes: float64(3), int64(3), object(6)
         memory usage: 436.2+ KB
```

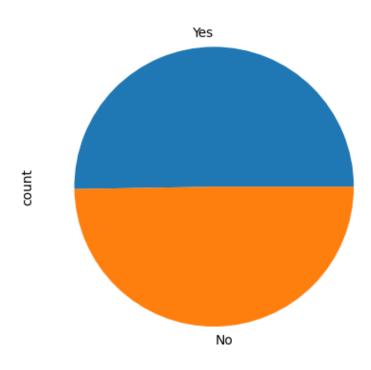
TechSupport:

Conclusion :-

- No Missing Values
- Value are almost divided equally

```
In [34]: df.TechSupport.value_counts()
Out[34]: TechSupport
    Yes    2338
    No    2313
    Name: count, dtype: int64

In [36]: df.TechSupport.value_counts().plot(kind='pie')
Out[36]: <Axes: ylabel='count'>
```



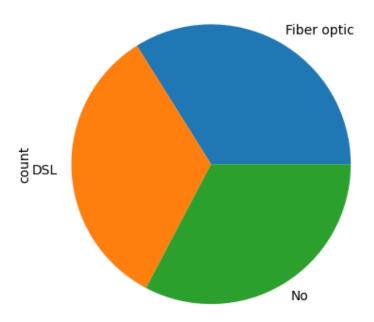
InternetService:

Conclusion:-

- No Missing Values
- Value are almost divided equally

```
In [38]: df.InternetService.value_counts().plot(kind = 'pie')
```

Out[38]: <Axes: ylabel='count'>



In [39]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4651 entries, 0 to 4650
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype	
0	CustomerID	4651 non-null	int64	
1	Age	4651 non-null	float64	
2	Gender	4651 non-null	object	
3	ContractType_In_days	4651 non-null	int64	
4	MonthlyCharges	4651 non-null	float64	
5	TotalCharges	4651 non-null	float64	
6	TechSupport	4651 non-null	object	
7	InternetService	4651 non-null	object	
8	Tenure	4651 non-null	int64	
9	PaperlessBilling	4651 non-null	object	
10	PaymentMethod	4651 non-null	object	
11	Churn	4651 non-null	object	
<pre>dtypes: float64(3), int64(3), object(6)</pre>				

memory usage: 436.2+ KB

PaperlessBilling:

`Conclusion`:-

- No Missing Values
- Value are almost divided equally

```
In [40]: df.PaperlessBilling.value_counts()
```

Out[40]: PaperlessBilling

No 2366 Yes 2285

Name: count, dtype: int64

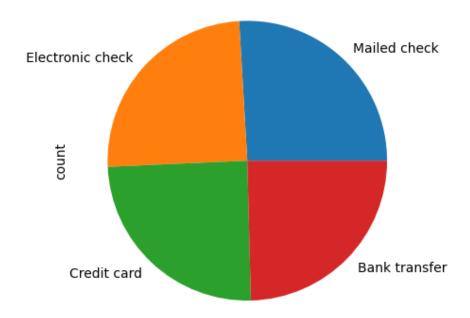
PaymentMethod:

Conclusion :-

- No Missing Values
- Value are almost divided equally

```
In [42]: df.PaymentMethod.value_counts().plot(kind='pie')
```

Out[42]: <Axes: ylabel='count'>



Churn:

Conclusion :-

- No Missing Values
- Value are almost divided equally

```
In [43]: df.Churn.value_counts()
Out[43]: Churn
    Yes    2374
    No    2277
    Name: count, dtype: int64
```

Univariate Analysis on Numerical columns

```
In [44]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 4651 entries, 0 to 4650
         Data columns (total 12 columns):
          #
              Column
                                    Non-Null Count
                                                    Dtype
              -----
                                     -----
          0
              CustomerID
                                                     int64
                                     4651 non-null
          1
                                    4651 non-null
                                                     float64
              Age
          2
              Gender
                                    4651 non-null
                                                     object
          3
              ContractType_In_days
                                    4651 non-null
                                                     int64
          4
              MonthlyCharges
                                    4651 non-null
                                                     float64
          5
              TotalCharges
                                    4651 non-null
                                                     float64
              TechSupport
                                    4651 non-null
                                                     object
          7
              InternetService
                                                     object
                                    4651 non-null
          8
              Tenure
                                    4651 non-null
                                                     int64
          9
              PaperlessBilling
                                    4651 non-null
                                                     object
          10 PaymentMethod
                                    4651 non-null
                                                     object
          11 Churn
                                                     object
                                    4651 non-null
         dtypes: float64(3), int64(3), object(6)
         memory usage: 436.2+ KB
```

Age:

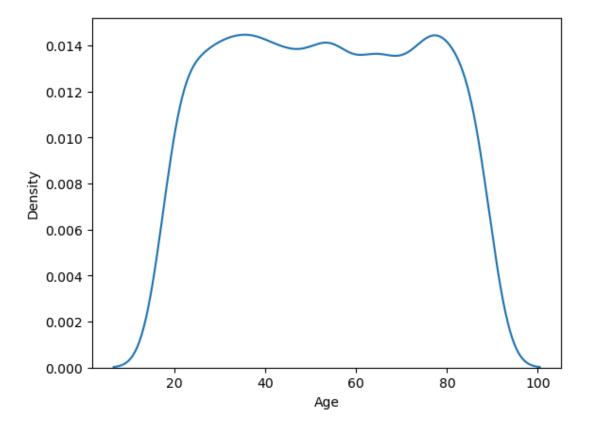
Conclusion :-

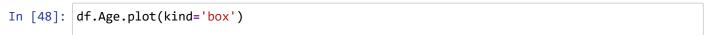
- No Missing Values
- No Outlier Found

```
In [45]: df.Age.describe()
Out[45]: count
                   4651.000000
         mean
                     53.321221
          std
                     20.679189
         min
                     18.000000
          25%
                     35.000000
          50%
                     53.000000
         75%
                     71.000000
         max
                     89.000000
         Name: Age, dtype: float64
```

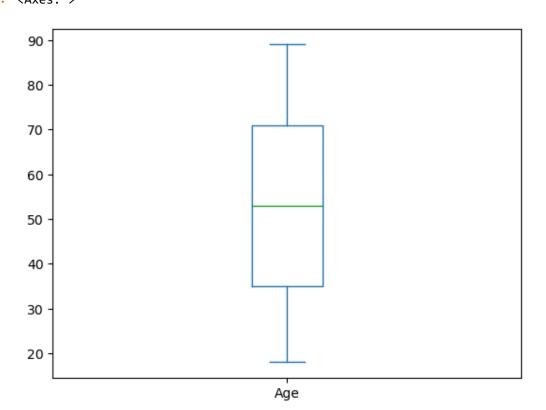
```
In [47]: sns.kdeplot(data=df,x='Age')
```

Out[47]: <Axes: xlabel='Age', ylabel='Density'>





Out[48]: <Axes: >



```
In [49]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4651 entries, 0 to 4650
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	CustomerID	4651 non-null	int64
1	Age	4651 non-null	float64
2	Gender	4651 non-null	object
3	ContractType_In_days	4651 non-null	int64
4	MonthlyCharges	4651 non-null	float64
5	TotalCharges	4651 non-null	float64
6	TechSupport	4651 non-null	object
7	InternetService	4651 non-null	object
8	Tenure	4651 non-null	int64
9	PaperlessBilling	4651 non-null	object
10	PaymentMethod	4651 non-null	object
11	Churn	4651 non-null	object
م ـ ـ ـ ـ ـ	aa. £1aa+c4/2\ :a+c4/	2) abiaat(C)	

dtypes: float64(3), int64(3), object(6)

memory usage: 436.2+ KB

MonthlyCharges:

Conclusion :-

- No Missing Values
- Outliers have been handled

```
In [50]: df.MonthlyCharges.describe()
```

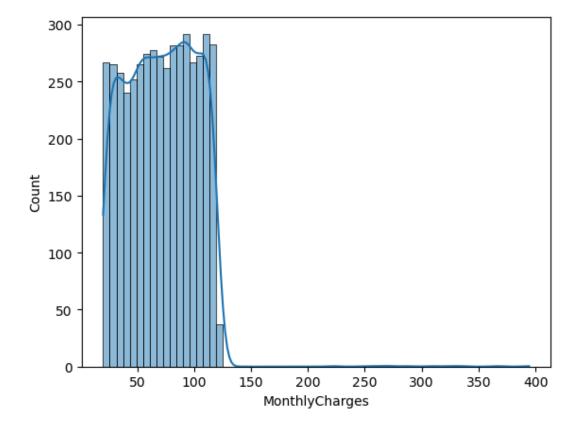
```
Out[50]: count
```

```
count 4651.000000
mean 71.586361
std 30.894229
min 20.030708
25% 46.904230
50% 71.630182
75% 96.053310
max 394.364052
```

Name: MonthlyCharges, dtype: float64

```
In [52]: sns.histplot(data=df,x='MonthlyCharges',kde=True)
```

```
Out[52]: <Axes: xlabel='MonthlyCharges', ylabel='Count'>
```



```
In [53]: df.MonthlyCharges.skew()
```

Out[53]: 0.944190282237162

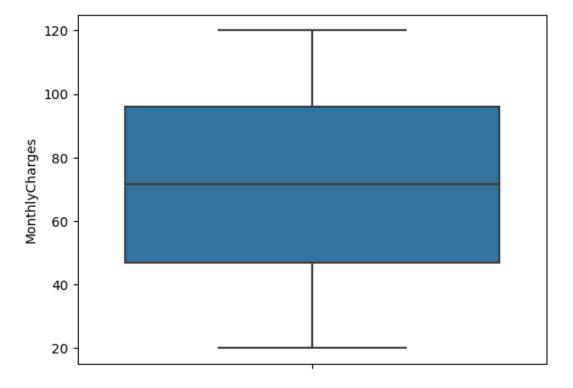
```
In [61]: df = df[~(df.MonthlyCharges>200)]
```

In [62]: df.MonthlyCharges.skew()

Out[62]: -0.05023282989140127

```
In [65]: sns.boxplot(data=df,y='MonthlyCharges')
```

```
Out[65]: <Axes: ylabel='MonthlyCharges'>
```



```
In [66]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
Index: 4641 entries, 0 to 4650
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	CustomerID	4641 non-null	int64
1	Age	4641 non-null	float64
2	Gender	4641 non-null	object
3	ContractType_In_days	4641 non-null	int64
4	MonthlyCharges	4641 non-null	float64
5	TotalCharges	4641 non-null	float64
6	TechSupport	4641 non-null	object
7	InternetService	4641 non-null	object
8	Tenure	4641 non-null	int64
9	PaperlessBilling	4641 non-null	object
10	PaymentMethod	4641 non-null	object
11	Churn	4641 non-null	object
44	an float(4/2) int(4/	2) abiaat(C)	-

dtypes: float64(3), int64(3), object(6)

memory usage: 471.4+ KB

TotalCharges:

Conclusion :-

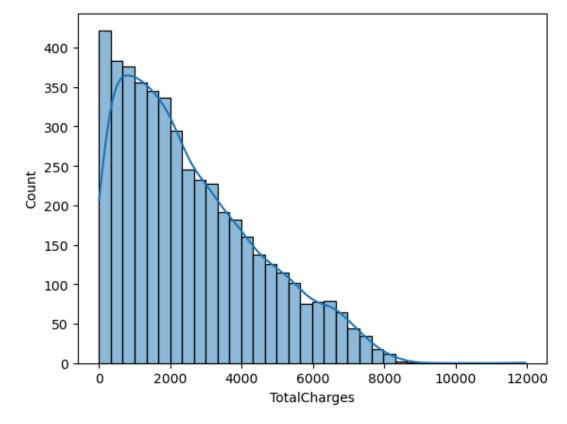
- No Missing Values
- Around 250 rows have droped as it contains the outlier values

```
df.TotalCharges.describe()
In [67]:
Out[67]: count
                    4641.000000
          mean
                    2547.948386
                    1926.446844
          std
          min
                       0.000000
          25%
                     981.664836
          50%
                    2106.499950
          75%
                    3788.770844
                   11943.264877
          max
          Name: TotalCharges, dtype: float64
In [68]:
         df.TotalCharges.skew()
```

Out[68]: 0.7841534300429074

In [70]: sns.histplot(data=df,x='TotalCharges',kde=True)

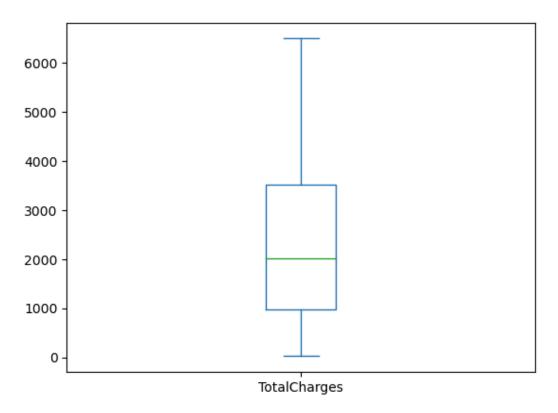
Out[70]: <Axes: xlabel='TotalCharges', ylabel='Count'>



```
In [78]: df = df[~(df.TotalCharges>6500)]
In [85]: df = df[~(df.TotalCharges==0)]
In [86]: df.TotalCharges.skew()
Out[86]: 0.6042191935647911
```

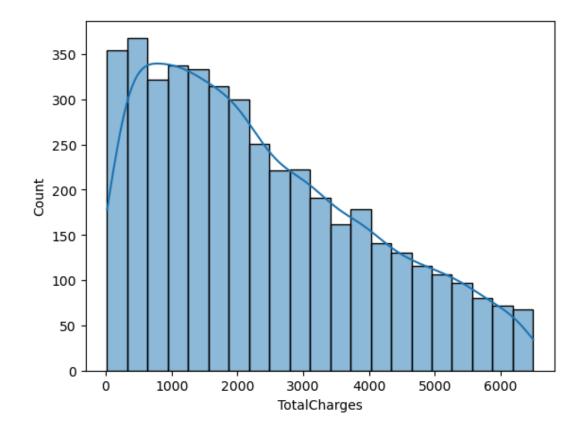
```
In [87]: df.TotalCharges.plot(kind='box')
```

Out[87]: <Axes: >



In [88]: sns.histplot(data=df,x='TotalCharges',kde=True)

Out[88]: <Axes: xlabel='TotalCharges', ylabel='Count'>



```
df.TotalCharges.describe()
In [89]:
Out[89]: count
                   4364.000000
         mean
                   2362.864874
                   1665.823190
          std
         min
                     20.099899
         25%
                    980.043353
         50%
                   2020.433565
          75%
                   3514.907060
                   6494.540110
         max
         Name: TotalCharges, dtype: float64
In [90]: df.info()
          <class 'pandas.core.frame.DataFrame'>
```

<class 'pandas.core.frame.DataFrame'>
Index: 4364 entries, 0 to 4649
Data columns (total 12 columns):

	\	/ -	
#	Column	Non-Null Count	Dtype
0	CustomerID	4364 non-null	int64
1	Age	4364 non-null	float64
2	Gender	4364 non-null	object
3	ContractType_In_days	4364 non-null	int64
4	MonthlyCharges	4364 non-null	float64
5	TotalCharges	4364 non-null	float64
6	TechSupport	4364 non-null	object
7	InternetService	4364 non-null	object
8	Tenure	4364 non-null	int64
9	PaperlessBilling	4364 non-null	object
10	PaymentMethod	4364 non-null	object
11	Churn	4364 non-null	object
d+vn	$ac \cdot float64(3) int64($	3) object(6)	

dtypes: float64(3), int64(3), object(6)

memory usage: 443.2+ KB

Tenure:

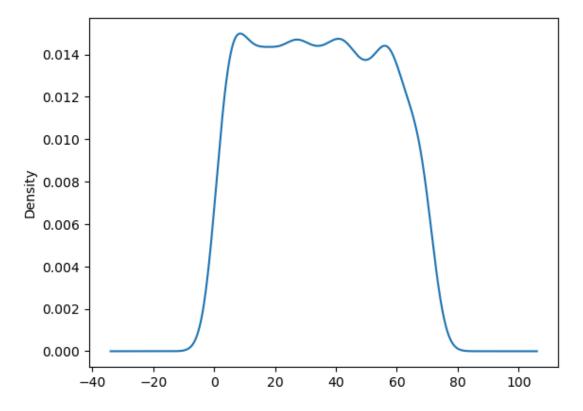
Conclusion:-

- No Missing Values
- Value are almost divided equally

```
In [105]: df.Tenure.describe()
Out[105]: count
                    4364.000000
          mean
                      34.683547
          std
                      20.125995
                       1.000000
          min
          25%
                      17.000000
          50%
                      34.000000
          75%
                      52.000000
                      71.000000
          Name: Tenure, dtype: float64
```

```
In [107]: df.Tenure.plot(kind='kde')
```

Out[107]: <Axes: ylabel='Density'>



In [108]: df.info()

<class 'pandas.core.frame.DataFrame'>
Index: 4364 entries, 0 to 4649
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	CustomerID	4364 non-null	int64
1	Age	4364 non-null	float64
2	Gender	4364 non-null	object
3	ContractType_In_days	4364 non-null	int64
4	MonthlyCharges	4364 non-null	float64
5	TotalCharges	4364 non-null	float64
6	TechSupport	4364 non-null	object
7	InternetService	4364 non-null	object
8	Tenure	4364 non-null	int64
9	PaperlessBilling	4364 non-null	object
10	PaymentMethod	4364 non-null	object
11	Churn	4364 non-null	object
dtyp	es: float64(3), int64(3), object(6)	-

memory usage: 443.2+ KB

```
In [109]: df
```

Out[109]:

	CustomerID	Age	Gender	ContractType_In_days	MonthlyCharges	TotalCharges	TechSupport	Intern
0	1083	79.0	Male	365	90.038513	3511.502019	No	
1	1117	60.0	Female	365	80.590894	2901.272196	No	
3	3833	84.0	Female	365	43.042067	1549.514395	No	
4	1976	69.0	Male	365	51.930032	2232.991377	No	
6	3132	49.0	Male	365	101.524194	913.717747	Yes	
4645	2133	39.0	Male	365	30.017101	210.119705	No	
4646	1514	54.0	Male	365	57.803077	462.424613	No	
4647	2716	45.0	Male	730	103.314530	826.516243	No	
4648	756	21.0	Female	730	103.105344	103.105344	Yes	
4649	3284	85.0	Male	730	36.907180	1660.823112	Yes	
4364 rows × 12 columns								
4								_ N

Bivariate Analysis

Categorical - Categorical

```
In [110]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
Index: 4364 entries, 0 to 4649
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	CustomerID	4364 non-null	int64
1	Age	4364 non-null	float64
2	Gender	4364 non-null	object
3	ContractType_In_days	4364 non-null	int64
4	MonthlyCharges	4364 non-null	float64
5	TotalCharges	4364 non-null	float64
6	TechSupport	4364 non-null	object
7	InternetService	4364 non-null	object
8	Tenure	4364 non-null	int64
9	PaperlessBilling	4364 non-null	object
10	PaymentMethod	4364 non-null	object
11	Churn	4364 non-null	object

dtypes: float64(3), int64(3), object(6)

memory usage: 443.2+ KB

Gender and TechSupport

--> Females Likes to support the tech

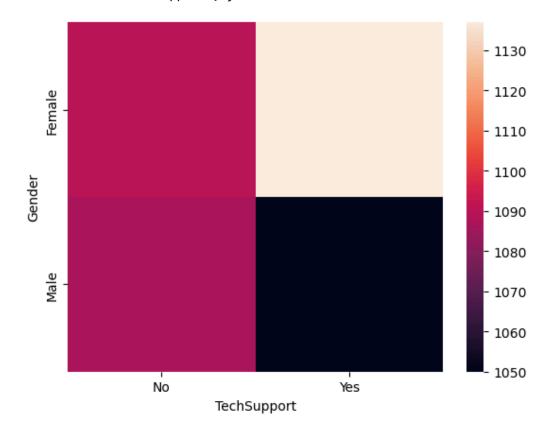
```
In [115]: pd.crosstab(df.Gender,df.TechSupport)
```

Out[115]:

TechSupport	No	Yes
Gender		
Female	1090	1137
Male	1087	1050

In [114]: sns.heatmap(pd.crosstab(df.Gender,df.TechSupport))

Out[114]: <Axes: xlabel='TechSupport', ylabel='Gender'>



Gender and InternetService

In [124]: pd.crosstab(df.Gender,df.InternetService)

Out[124]:

InternetService		DSL	Fiber optic	No
	Gender			
	Female	730	773	724
	Male	745	694	698

```
In [125]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
Index: 4364 entries, 0 to 4649
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	CustomerID	4364 non-null	int64
1	Age	4364 non-null	float64
2	Gender	4364 non-null	object
3	ContractType_In_days	4364 non-null	int64
4	MonthlyCharges	4364 non-null	float64
5	TotalCharges	4364 non-null	float64
6	TechSupport	4364 non-null	object
7	InternetService	4364 non-null	object
8	Tenure	4364 non-null	int64
9	PaperlessBilling	4364 non-null	object
10	PaymentMethod	4364 non-null	object
11	Churn	4364 non-null	object
4+,,,,	oc. float(1/2) int(1/	2) object(6)	-

dtypes: float64(3), int64(3), object(6)

memory usage: 443.2+ KB

Male

Gender and PaymentMethod, Paperless Billing

In [126]: pd.crosstab(df.Gender,df.PaymentMethod)

523

542

Out[126]:

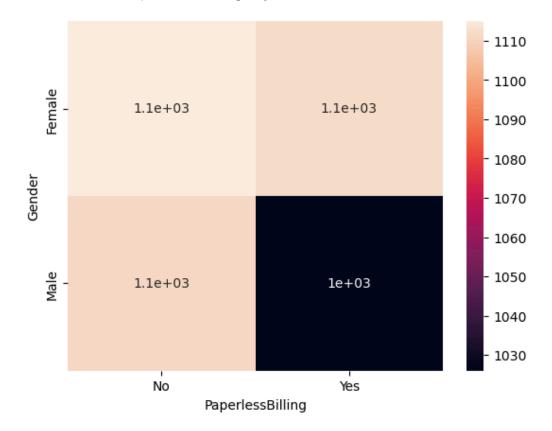
PaymentMethod	Bank transfer	Credit card	Electronic check	Mailed check
Gender				
Female	537	557	556	577

531

541

In [128]: sns.heatmap(pd.crosstab(df.Gender,df.PaperlessBilling),annot=True)

Out[128]: <Axes: xlabel='PaperlessBilling', ylabel='Gender'>



Churn and Gender

--> Generally females are not statisfied with the services provided by the company

In [155]: pd.crosstab(df.Churn,df.Gender)

Out[155]:

Gender	Female	Male	
Churn			
No	1092	1039	
Yes	1135	1098	

Numerical - Categorical

```
In [129]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
Index: 4364 entries, 0 to 4649
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype		
0	CustomerID	4364 non-null	int64		
1	Age	4364 non-null	float64		
2	Gender	4364 non-null	object		
3	ContractType_In_days	4364 non-null	int64		
4	MonthlyCharges	4364 non-null	float64		
5	TotalCharges	4364 non-null	float64		
6	TechSupport	4364 non-null	object		
7	InternetService	4364 non-null	object		
8	Tenure	4364 non-null	int64		
9	PaperlessBilling	4364 non-null	object		
10	PaymentMethod	4364 non-null	object		
11	Churn	4364 non-null	object		
dtypes: fleet64(2) int64(2) object(6)					

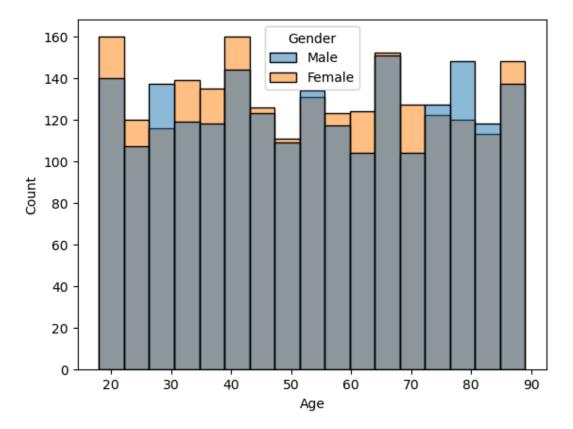
dtypes: float64(3), int64(3), object(6)

memory usage: 443.2+ KB

Gender and Age

```
In [137]: sns.histplot(data=df,x='Age',hue='Gender')
```

Out[137]: <Axes: xlabel='Age', ylabel='Count'>



```
In [133]: df.sample(5)
```

Out[133]:

	CustomerID	Age	Gender	ContractType_In_days	MonthlyCharges	TotalCharges	TechSupport	Intern
3607	4406	21.0	Male	30	89.551048	805.959429	Yes	
4179	2011	85.0	Male	730	91.927139	5055.992618	Yes	
4560	3453	46.0	Male	365	118.828859	5347.298652	No	
3544	2667	36.0	Female	30	47.225436	377.803488	No	
4110	1430	88.0	Female	365	22.864593	297.239711	Yes	
4								•

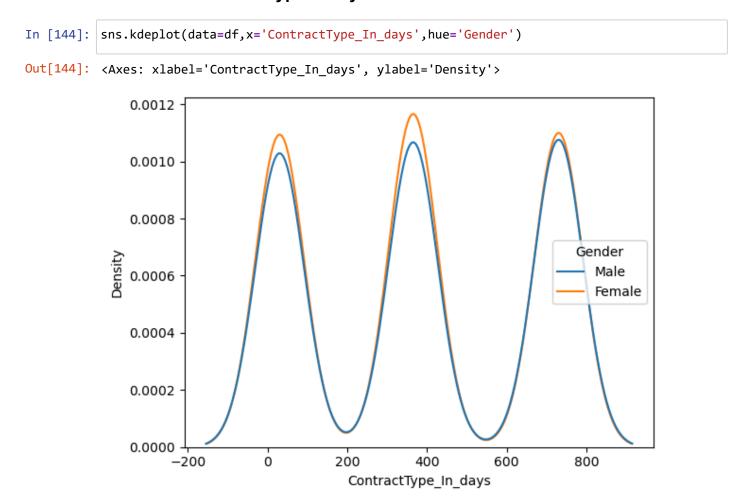
In [138]: | df.info()

<class 'pandas.core.frame.DataFrame'> Index: 4364 entries, 0 to 4649 Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype	
0	CustomerID	4364 non-null	int64	
1	Age	4364 non-null	float64	
2	Gender	4364 non-null	object	
3	ContractType_In_days	4364 non-null	int64	
4	MonthlyCharges	4364 non-null	float64	
5	TotalCharges	4364 non-null	float64	
6	TechSupport	4364 non-null	object	
7	InternetService	4364 non-null	object	
8	Tenure	4364 non-null	int64	
9	PaperlessBilling	4364 non-null	object	
10	PaymentMethod	4364 non-null	object	
11	Churn	object		
<pre>dtypes: float64(3), int64(3), object(6)</pre>				

memory usage: 443.2+ KB

Gender and Contracttype in days



TotalCharges and InternetService

In [150]: df[['InternetService','TotalCharges']].groupby('InternetService').mean()

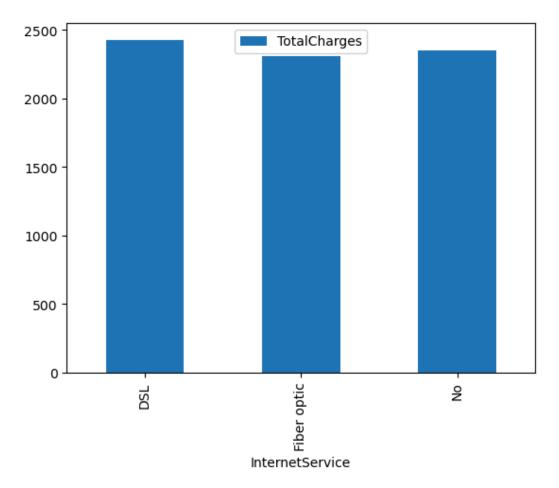
Out[150]:

TotalCharges

InternetService				
DSL	2427.036222			
Fiber optic	2308.772171			
No	2352.106266			

```
In [152]: df[['InternetService','TotalCharges']].groupby('InternetService').mean().plot(kind='ba
```

Out[152]: <Axes: xlabel='InternetService'>



In [153]: df.info()

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

Index: 4364 entries, 0 to 4649
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype		
0	CustomerID	4364 non-null	int64		
1	Age	4364 non-null	float64		
2	Gender	4364 non-null	object		
3	ContractType_In_days	4364 non-null	int64		
4	MonthlyCharges	4364 non-null	float64		
5	TotalCharges	4364 non-null	float64		
6	TechSupport	4364 non-null	object		
7	InternetService	4364 non-null	object		
8	Tenure	4364 non-null	int64		
9	PaperlessBilling	4364 non-null	object		
10	PaymentMethod	4364 non-null	object		
11	Churn	4364 non-null	object		
dtypes: float64(3), int64(3), object(6)					

memory usage: 443.2+ KB

Churn and MonthlyCharges

```
In [161]: df.Churn.value_counts()
Out[161]: Churn
           Yes
                  2233
                  2131
           No
           Name: count, dtype: int64
In [162]: |df[['Churn', 'MonthlyCharges']].groupby('Churn').mean()
Out[162]:
                  MonthlyCharges
            Churn
              No
                       69.526712
              Yes
                       69.070298
```

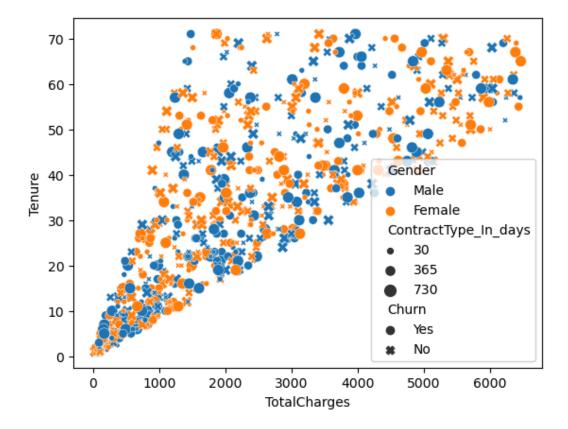
Numerical and Numerical

```
In [163]: df.info()
          <class 'pandas.core.frame.DataFrame'>
          Index: 4364 entries, 0 to 4649
          Data columns (total 12 columns):
               Column
                                      Non-Null Count
                                                      Dtype
               CustomerID
           0
                                      4364 non-null
                                                      int64
           1
               Age
                                      4364 non-null
                                                      float64
           2
               Gender
                                      4364 non-null
                                                      object
               ContractType_In_days 4364 non-null
                                                      int64
           4
               MonthlyCharges
                                      4364 non-null
                                                      float64
           5
               TotalCharges
                                      4364 non-null
                                                      float64
           6
               TechSupport
                                      4364 non-null
                                                      object
           7
               InternetService
                                      4364 non-null
                                                      object
           8
               Tenure
                                      4364 non-null
                                                      int64
           9
               PaperlessBilling
                                      4364 non-null
                                                      object
           10 PaymentMethod
                                      4364 non-null
                                                      object
           11 Churn
                                      4364 non-null
                                                      object
          dtypes: float64(3), int64(3), object(6)
```

memory usage: 443.2+ KB

```
In [174]: sns.scatterplot(data=df.sample(700),x='TotalCharges',y='Tenure',hue='Gender',style='Ch
```

Out[174]: <Axes: xlabel='TotalCharges', ylabel='Tenure'>



```
In [175]: data = df
```

In [176]: data.to_csv('EDA_Customer.csv')