TELECOM-CUSTOMER-CHURN-EDA-PROJECT_VIVEK_CHAUHAN

```
In [1]: # import necessary Libraries to work with dataset
    import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
In [2]: data = pd.read_csv("C:/Users/VIVEK CHAUHAN/Desktop/eda-projects (1)/7-eda-project/Telco-Customer-Churn.csv")
```

7590- VHVEG 5575- GNVDE 3668- QPYBK 7795- CFOCW 9237- HQITU	Female Male Male Female	0 0 0	Yes No No No	No No No	1 34 2 45	Yes Yes No	No phone service No No No phone service	DSL DSL DSL	Yes Yes
GNVDE 3668- QPYBK 7795- CFOCW 9237- HQITU	Male Male	0	No No	No	2	Yes	No No phone	DSL	Yes
QPYBK 7795- CFOCW 9237- HQITU	Male	0	No				No phone		
9237- HQITU				No	45	No		DSL	Yes
HQITU	Female	0	No						
				No	2	Yes	No	Fiber optic	No
•••									
840-RESVB	Male	0	Yes	Yes	24	Yes	Yes	DSL	Yes
2234- XADUH	Female	0	Yes	Yes	72	Yes	Yes	Fiber optic	No
4801-JZAZL	Female	0	Yes	Yes	11	No	No phone service	DSL	Yes
8361- LTMKD	Male	1	Yes	No	4	Yes	Yes	Fiber optic	No
3186-AJIEK	Male	0	No	No	66	Yes	No	Fiber optic	Yes
3	2234- XADUH 801-JZAZL 8361- LTMKD	2234- XADUH Female 801-JZAZL Female 8361- LTMKD Male	2234- XADUH Female 0 801-JZAZL Female 0 8361- LTMKD Male 1 1186-AJIEK Male 0	2234- XADUH Female 0 Yes 801-JZAZL Female 0 Yes 8361- LTMKD Male 1 Yes 8186-AJIEK Male 0 No	2234- XADUH Female 0 Yes Yes 801-JZAZL Female 0 Yes Yes 8361- LTMKD Male 1 Yes No	2234- XADUH Female 0 Yes Yes 72 801-JZAZL Female 0 Yes Yes 11 8361- LTMKD Male 1 Yes No 4 8186-AJIEK Male 0 No No 66	2234- XADUH Female 0 Yes Yes 72 Yes 801-JZAZL Female 0 Yes Yes 11 No 8361- LTMKD Male 1 Yes No 4 Yes 8186-AJIEK Male 0 No No 66 Yes	2234- XADUHFemale0YesYes72YesYes801-JZAZLFemale0YesYes11NoNo phone service8361- LTMKDMale1YesNo4YesYes8186-AJIEKMale0NoNo66YesNo	2234- XADUH Female 0 Yes Yes 72 Yes Yes Fiber optic 801-JZAZL Female 0 Yes Yes 11 No No phone service DSL 8361- LTMKD Male 1 Yes No 4 Yes Yes Fiber optic 1186-AJIEK Male 0 No No 66 Yes No Fiber optic

Out[4]:		customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	•••	Dev
	0	7590- VHVEG	Female	0	Yes	No	1	No	No phone service	DSL	No		
	1	5575- GNVDE	Male	0	No	No	34	Yes	No	DSL	Yes		
	2	3668- QPYBK	Male	0	No	No	2	Yes	No	DSL	Yes		
	3	7795- CFOCW	Male	0	No	No	45	No	No phone service	DSL	Yes		
	4	9237- HQITU	Female	0	No	No	2	Yes	No	Fiber optic	No		
	5	9305- CDSKC	Female	0	No	No	8	Yes	Yes	Fiber optic	No		
	6	1452-KIOVK	Male	0	No	Yes	22	Yes	Yes	Fiber optic	No		
	7	6713- OKOMC	Female	0	No	No	10	No	No phone service	DSL	Yes		
	8	7892- POOKP	Female	0	Yes	No	28	Yes	Yes	Fiber optic	No		
	9	6388- TABGU	Male	0	No	Yes	62	Yes	No	DSL	Yes		

10 rows × 21 columns

→

In [5]: # Last 10 data from the dataset
 data.tail(10)

Out[5]:		customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	•••
	7033	9767-FFLEM	Male	0	No	No	38	Yes	No	Fiber optic	No	
	7034	0639- TSIQW	Female	0	No	No	67	Yes	Yes	Fiber optic	Yes	
	7035	8456- QDAVC	Male	0	No	No	19	Yes	No	Fiber optic	No	
	7036	7750- EYXWZ	Female	0	No	No	12	No	No phone service	DSL	No	
	7037	2569- WGERO	Female	0	No	No	72	Yes	No	No	No internet service	
	7038	6840-RESVB	Male	0	Yes	Yes	24	Yes	Yes	DSL	Yes	
	7039	2234- XADUH	Female	0	Yes	Yes	72	Yes	Yes	Fiber optic	No	•••
	7040	4801-JZAZL	Female	0	Yes	Yes	11	No	No phone service	DSL	Yes	
	7041	8361- LTMKD	Male	1	Yes	No	4	Yes	Yes	Fiber optic	No	
	7042	3186-AJIEK	Male	0	No	No	66	Yes	No	Fiber optic	Yes	

10 rows × 21 columns

In [6]: # checking the data types of all the columns

```
data.dtypes
Out[6]: customerID
                             object
                             object
        gender
                              int64
        SeniorCitizen
        Partner
                             object
        Dependents
                             object
        tenure
                              int64
                             object
        PhoneService
        MultipleLines
                             object
        InternetService
                             object
        OnlineSecurity
                             object
        OnlineBackup
                             object
                             object
        DeviceProtection
                             object
        TechSupport
        StreamingTV
                             object
                             object
        StreamingMovies
        Contract
                             object
        PaperlessBilling
                             object
        PaymentMethod
                             object
                            float64
        MonthlyCharges
                             object
        TotalCharges
        Churn
                             object
        dtype: object
        # info about our dataset
        data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 21 columns):
```

Data	COTUMNIS (COCAT ZI	corumns).								
#	Column	Non-Null Count	Dtype							
0	customerID	7043 non-null	object							
1	gender	7043 non-null	object							
2	SeniorCitizen	7043 non-null	int64							
3	Partner	7043 non-null	object							
4	Dependents	7043 non-null	object							
5	tenure	7043 non-null	int64							
6	PhoneService	7043 non-null	object							
7	MultipleLines	7043 non-null	object							
8	InternetService	7043 non-null	object							
9	OnlineSecurity	7043 non-null	object							
10	OnlineBackup	7043 non-null	object							
11	DeviceProtection	7043 non-null	object							
12	TechSupport	7043 non-null	object							
13	StreamingTV	7043 non-null	object							
14	StreamingMovies	7043 non-null	object							
15	Contract	7043 non-null	object							
16	PaperlessBilling	7043 non-null	object							
17	PaymentMethod	7043 non-null	object							
18	MonthlyCharges	7043 non-null	float64							
19	TotalCharges	7043 non-null	object							
20	Churn	7043 non-null	object							
dtype	es: float64(1), int	t64(2), object(18	3)							
momony usago. 1 1, MD										

memory usage: 1.1+ MB

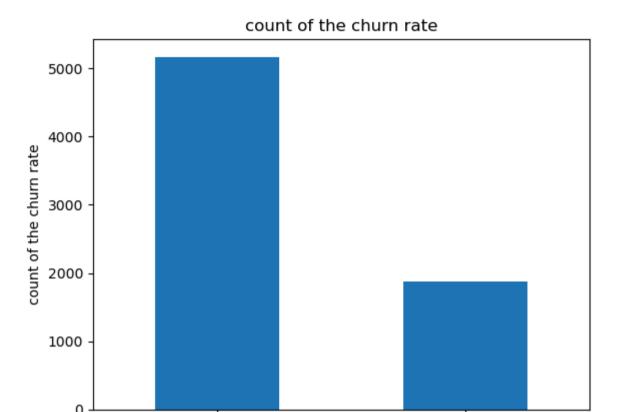
```
In [8]: # get the satistics about our dataser
        data.describe()
```

Out[8]:		SeniorCitizen	tenure	MonthlyCharges		
	count	7043.000000	7043.000000	7043.000000		
	mean	0.162147	32.371149	64.761692		
	std	0.368612	24.559481	30.090047		
	min	0.000000	0.000000	18.250000		
	25%	0.000000	9.000000	35.500000		
	50%	0.000000	29.000000	70.350000		
	75 %	0.000000	55.000000	89.850000		
	max	1.000000	72.000000	118.750000		

```
In [9]: # churn rate

data["Churn"].value_counts().plot(kind="bar")
plt.xlabel("churn yes and no")
plt.ylabel("count of the churn rate")
plt.title("count of the churn rate")
```

Out[9]: Text(0.5, 1.0, 'count of the churn rate')



churn yes and no

```
In [10]: # churn counts
    data["Churn"].value_counts()

Out[10]: Churn
    No    5174
    Yes    1869
    Name: count, dtype: int64

In [11]: # percentage of the churn rate
    100 * data["Churn"].value_counts() / len(data["Churn"])
```

Yes

```
Out[11]: Churn
                73.463013
         No
                26,536987
         Yes
         Name: count, dtype: float64
In [12]: # check our dataset and take random column has any nan or empty values or not
         data["Churn"].hasnans
Out[12]: False
In [13]: # minimum tenures
         data["tenure"].min()
Out[13]: 0
In [14]: # max tenures
         data["tenure"].max()
Out[14]: 72
In [15]: # average tenures
         data["tenure"].mean()
Out[15]: 32.37114865824223
In [16]: # most common tenures
         data["tenure"].mode()
Out[16]: 0 1
         Name: tenure, dtype: int64
In [17]: # most common internet services so we can identify which service is public prefer
         data["InternetService"].mode()
```

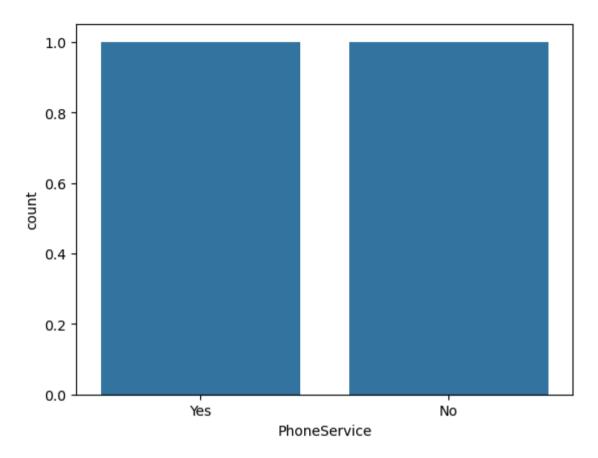
```
Out[17]: 0
              Fiber optic
         Name: InternetService, dtype: object
In [18]: # most of the customer which type of contracts have
         data["Contract"].mode()
              Month-to-month
Out[18]: 0
         Name: Contract, dtype: object
In [19]: # which type of payment method that customer used
         data["PaymentMethod"].mode()
Out[19]: 0
               Electronic check
         Name: PaymentMethod, dtype: object
In [20]: # Let's count the customers genders
         data["gender"].value counts()
Out[20]: gender
          Male
                    3555
          Female
                   3488
         Name: count, dtype: int64
In [21]: # Let's count the partner
         data["Partner"].value counts()
Out[21]: Partner
                 3641
          No
                 3402
          Yes
         Name: count, dtype: int64
In [22]: # Let's count the Dependents
         data["Dependents"].value_counts()
```

```
Out[22]: Dependents
No 4933
Yes 2110
Name: count, dtype: int64

In [23]: # Let's count the total tenure
data["tenure"].sum()

Out[23]: 227990

In [24]: # Let's count the phoneservices
a = data.PhoneService.value_counts()
sns.countplot(data=a)
plt.show()
```



```
Out[26]: InternetService
         Fiber optic
                         3096
         DSL
                         2421
          No
                         1526
         Name: count, dtype: int64
In [27]: # let's count the onlinesecurity services
         data.OnlineSecurity.value counts()
Out[27]: OnlineSecurity
                                 3498
         No
         Yes
                                 2019
         No internet service
                                 1526
         Name: count, dtype: int64
In [28]: # let's count the onlinebackup services
         data.OnlineBackup.value counts()
Out[28]: OnlineBackup
         No
                                 3088
         Yes
                                 2429
         No internet service
                                 1526
         Name: count, dtype: int64
In [29]: # Let's count the deviceprotection services
         data.DeviceProtection.value_counts()
Out[29]: DeviceProtection
                                 3095
         No
         Yes
                                 2422
         No internet service
                                 1526
         Name: count, dtype: int64
In [30]: # Let's count the TechSupport services
         data.TechSupport.value counts()
```

```
Out[30]: TechSupport
         No
                                 3473
                                 2044
         Yes
                                1526
         No internet service
         Name: count, dtype: int64
In [31]: # Let's count the StreamingTV services
         data.StreamingTV.value counts()
Out[31]: StreamingTV
         No
                                 2810
                                 2707
         Yes
         No internet service
                                1526
         Name: count, dtype: int64
In [32]: # Let's count the StreamingMovies services
         data.StreamingMovies.value counts()
Out[32]: StreamingMovies
         No
                                 2785
         Yes
                                 2732
         No internet service
                                1526
         Name: count, dtype: int64
In [33]: # let's count the Contract services
         data.Contract.value_counts()
Out[33]: Contract
         Month-to-month
                           3875
         Two year
                           1695
         One year
                           1473
         Name: count, dtype: int64
In [34]: # let's count the PaperlessBilling services
         data.PaperlessBilling.value counts()
```

```
Out[34]: PaperlessBilling
          Yes
                 4171
                 2872
          No
         Name: count, dtype: int64
In [35]: # Let's count the PaymentMethod services
         data.PaymentMethod.value counts()
Out[35]: PaymentMethod
         Electronic check
                                      2365
          Mailed check
                                      1612
         Bank transfer (automatic)
                                      1544
         Credit card (automatic)
                                      1522
         Name: count, dtype: int64
In [36]: # let's count the MonthlyCharges services
         data.MonthlyCharges.value counts()
Out[36]: MonthlyCharges
          20.05
                   61
         19.85
                   45
          19.95
                   44
         19.90
                   44
          20.00
                   43
          23.65
                    1
          114.70
                    1
         43.65
                    1
         87.80
                     1
         78.70
         Name: count, Length: 1585, dtype: int64
In [37]: # minimum MonthlyCharges
         data.MonthlyCharges.min()
Out[37]: 18.25
```

```
In [38]: # maximum MonthlyCharges
         data.MonthlyCharges.max()
Out[38]: 118.75
In [39]: # average mothlycharge()
         data.MonthlyCharges.mean()
Out[39]: 64.76169246059918
In [40]: # minimum TotalCharges
         data.TotalCharges.min()
Out[40]: ''
In [41]: # maximum TotalCharges
         data.TotalCharges.max()
Out[41]: '999.9'
In [42]: # check the type of totalcharges column data
         data.TotalCharges.dtype
Out[42]: dtype('0')
In [43]: # Let's count the churn customer
         data.Churn.value_counts()
Out[43]: Churn
                 5174
                1869
         Yes
         Name: count, dtype: int64
```

In [44]: # to check if the dataset has any blank or empty cell or values still present
data.isna()

Out[44]:

:		customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	•••
	0	False	False	False	False	False	False	False	False	False	False	
	1	False	False	False	False	False	False	False	False	False	False	
	2	False	False	False	False	False	False	False	False	False	False	
	3	False	False	False	False	False	False	False	False	False	False	
	4	False	False	False	False	False	False	False	False	False	False	
	•••											
7	038	False	False	False	False	False	False	False	False	False	False	
7	039	False	False	False	False	False	False	False	False	False	False	
7	040	False	False	False	False	False	False	False	False	False	False	
7	041	False	False	False	False	False	False	False	False	False	False	
7	042	False	False	False	False	False	False	False	False	False	False	

7043 rows × 21 columns

```
In [45]: # Let's find out the customer id whose minimum tenures
a = data.tenure.min()
b = data["customerID"][data["tenure"]==a]
b
```

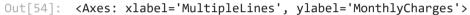
```
Out[45]: 488
                 4472-LVYGI
          753
                 3115-CZMZD
                 5709-LV0EQ
          936
         1082
                 4367-NUYA0
                 1371-DWPAZ
          1340
                 7644-0MVMY
          3331
         3826
                 3213-VV0LG
                 2520-SGTTA
          4380
                 2923-ARZLG
          5218
                 4075-WKNIU
          6670
         6754
                 2775-SEFEE
         Name: customerID, dtype: object
In [46]: # Let's find out the customer id, gender whose maximum tenures
         a = data.tenure.max()
         b = data[["customerID", "gender"]][data["tenure"]==a]
```

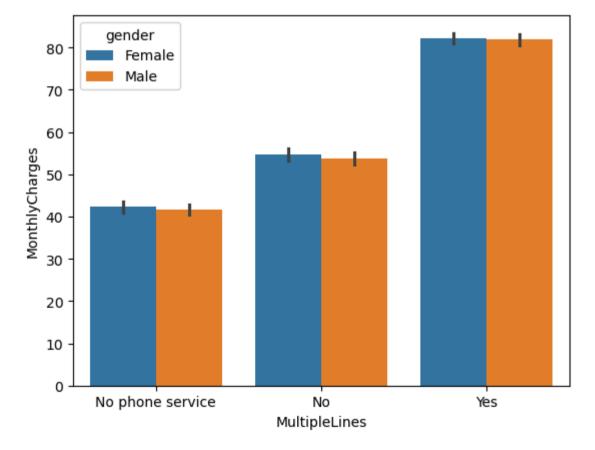
```
Out[46]:
                customerID gender
           28
                 5248-YGIJN
                              Male
                6234-RAAPL Female
                5954-BDFSG Female
                0526-SXDJP
                              Male
                 9848-JQJTX
                              Male
         6982
                8468-FZTOE Female
                2274-XUATA
         7007
                              Male
         7022
               7203-OYKCT
                              Male
         7037 2569-WGERO Female
         7039 2234-XADUH Female
        362 rows × 2 columns
In [47]: # Let's find out the min monthlycharges customer id, gender
         a = data.MonthlyCharges.min()
         b = data[["customerID", "gender"]][data["MonthlyCharges"]==a]
Out[47]:
               customerID gender
         3719 6823-SIDFQ
                             Male
In [48]: # Let's find out the max monthlycharges customer id,gender
         a = data.MonthlyCharges.max()
```

```
b = data[["customerID", "gender"]][data["MonthlyCharges"]==a]
Out[48]:
                customerID gender
         4586 7569-NMZYQ Female
In [49]: # let's find out the min TotalCharges customer id, gender
         a = data.TotalCharges.min()
         b = data[["customerID","gender"]][data["TotalCharges"]==a]
         b
Out[49]:
                 customerID gender
          488
                 4472-LVYGI Female
          753 3115-CZMZD
                              Male
                5709-LVOEQ Female
          936
         1082 4367-NUYAO
                              Male
               1371-DWPAZ Female
         1340
         3331 7644-OMVMY
                              Male
                3213-VVOLG
         3826
                              Male
         4380
                 2520-SGTTA Female
                 2923-ARZLG
         5218
                              Male
                4075-WKNIU
         6670
                            Female
         6754
                 2775-SEFEE
                              Male
In [50]: # Let's find out the max TotalCharges customer id, gender
         a = data.TotalCharges.max()
```

```
b = data[["customerID", "gender"]][data["TotalCharges"]==a]
Out[50]:
               customerID gender
          2845 9093-FPDLG Female
In [51]: # Let's count the gender and customer-id who already churned means yes
         b = data["gender"] [data.Churn == "Yes"]
         b.value counts()
Out[51]: gender
          Female
                   939
          Male
                   930
         Name: count, dtype: int64
In [52]: # Let's count the gender and customer-id who not churned means no
         b = data["gender"] [data.Churn == "No"]
         b.value counts()
Out[52]: gender
          Male
                    2625
          Female
                   2549
         Name: count, dtype: int64
In [53]: # change the datatype of TotalCharges column
         data['TotalCharges'] = data['TotalCharges'].replace(r'^\s*$', np.nan, regex=True)
         data['TotalCharges'] = data['TotalCharges'].astype(float)
In [77]: # count the null values in data['TotalCharges'] column
         data['TotalCharges'].isnull().sum()
Out[77]: 11
In [80]: # fill the null cells with means
```

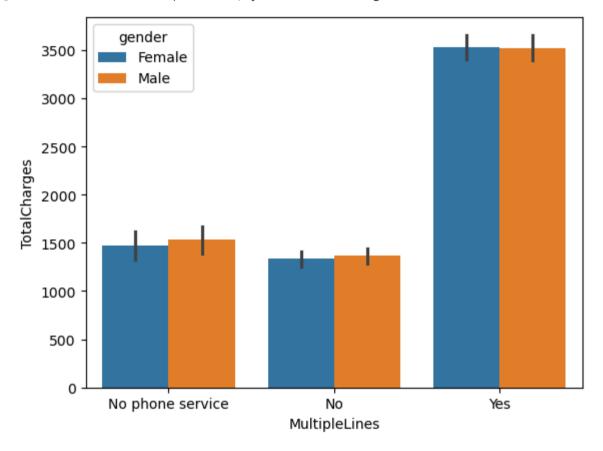
```
avg_totalcharges = data['TotalCharges'].mean()
         data['TotalCharges'] = data['TotalCharges'].fillna(avg totalcharges)
In [81]: # count the null values in data['TotalCharges'] column
         data['TotalCharges'].isnull().sum()
Out[81]: 0
        # MultipleLines wise monthly charges
         sns.barplot(x="MultipleLines",y="MonthlyCharges",data=data,hue="gender")
```





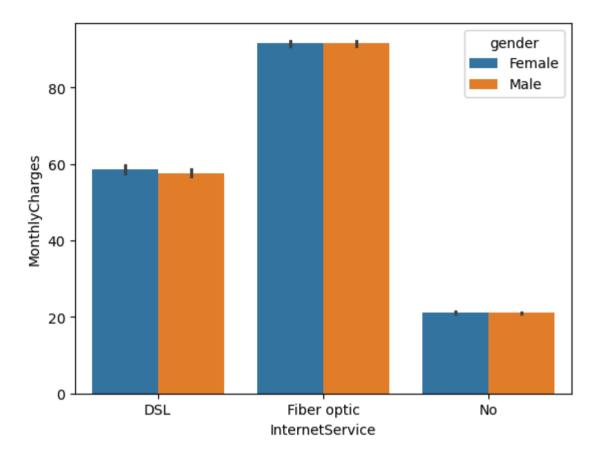
```
In [55]: # MultipleLines wise total charges
sns.barplot(x="MultipleLines",y="TotalCharges",data=data,hue="gender")
```

Out[55]: <Axes: xlabel='MultipleLines', ylabel='TotalCharges'>



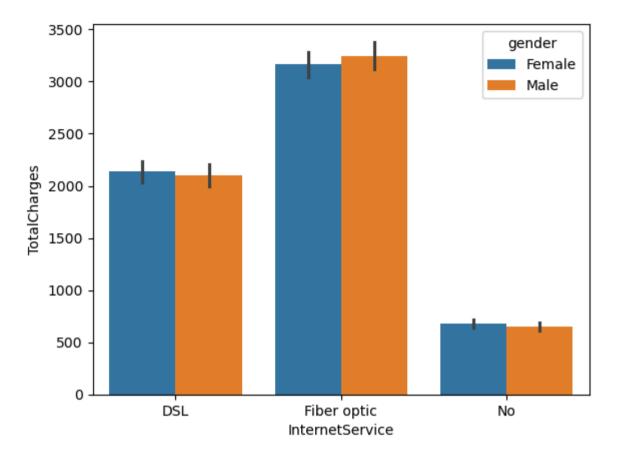
```
In [56]: # InternetService wise MonthlyCharges
sns.barplot(x="InternetService",y="MonthlyCharges",data=data,hue="gender")
```

Out[56]: <Axes: xlabel='InternetService', ylabel='MonthlyCharges'>



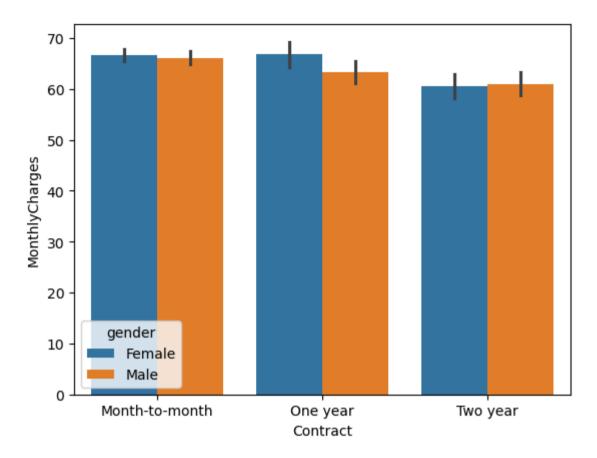
```
In [57]: # InternetService wise total charges
sns.barplot(x="InternetService",y="TotalCharges",data=data,hue="gender")
```

Out[57]: <Axes: xlabel='InternetService', ylabel='TotalCharges'>



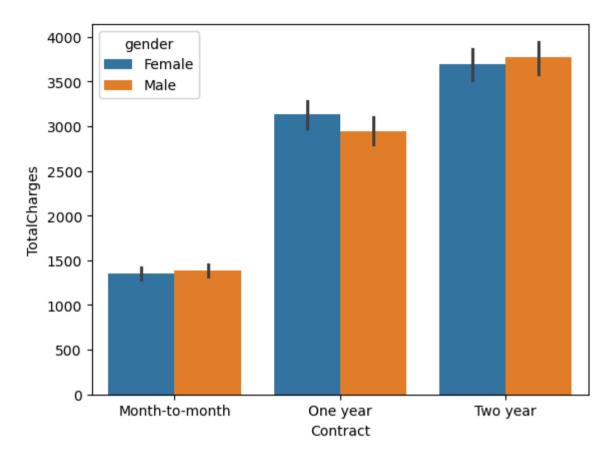
```
In [58]: # Contract wise MonthlyCharges
sns.barplot(x="Contract",y="MonthlyCharges",data=data,hue="gender")
```

Out[58]: <Axes: xlabel='Contract', ylabel='MonthlyCharges'>

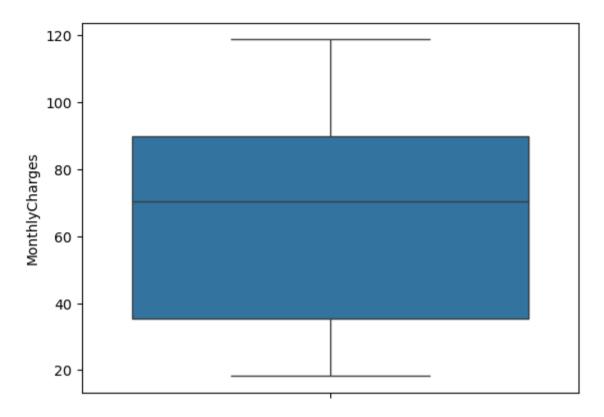


```
In [59]: # Contract wise TotalCharges
sns.barplot(x="Contract",y="TotalCharges",data=data,hue="gender")
```

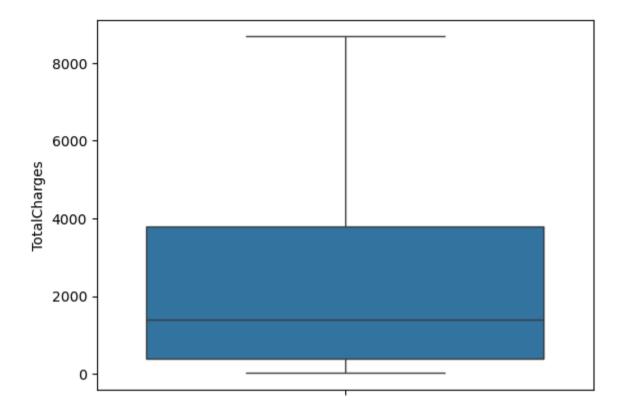
Out[59]: <Axes: xlabel='Contract', ylabel='TotalCharges'>



In [60]: # let's check is there any outliers in our dataset
sns.boxplot(y="MonthlyCharges",data=data)
plt.show()

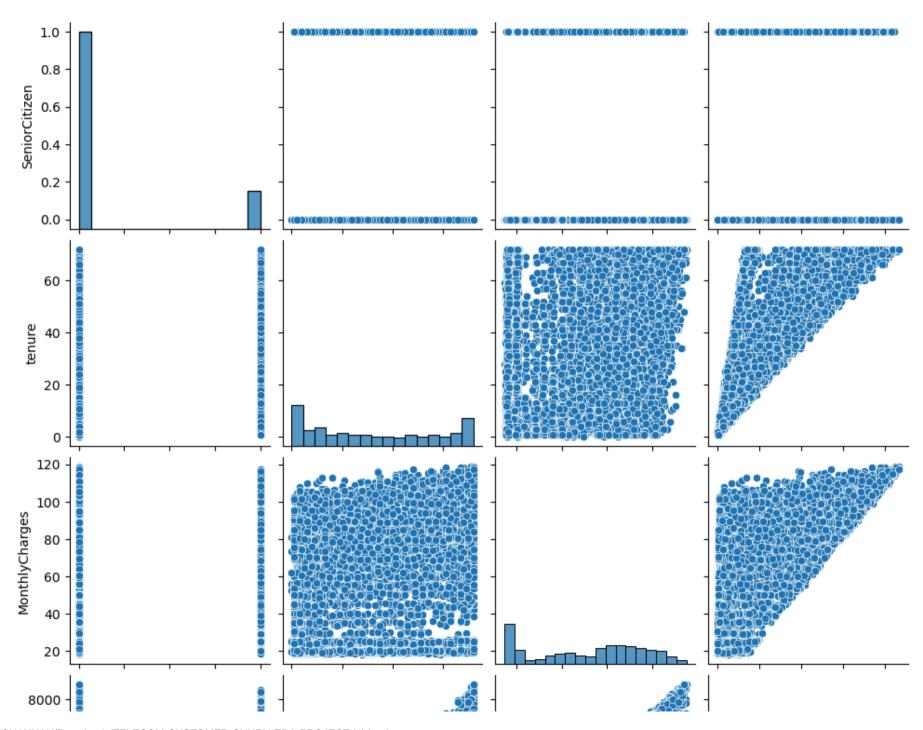


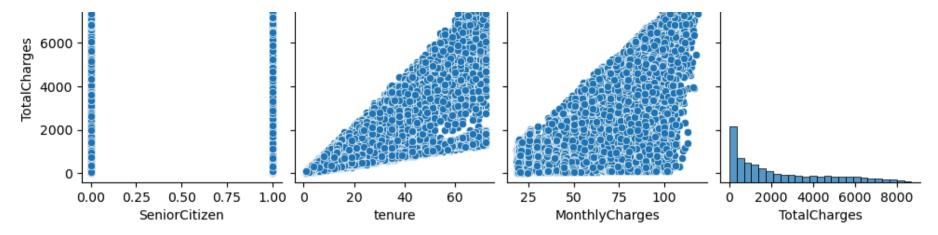
```
In [61]: # Let's check is there any outliers in our dataset
sns.boxplot(y="TotalCharges",data=data)
plt.show()
```



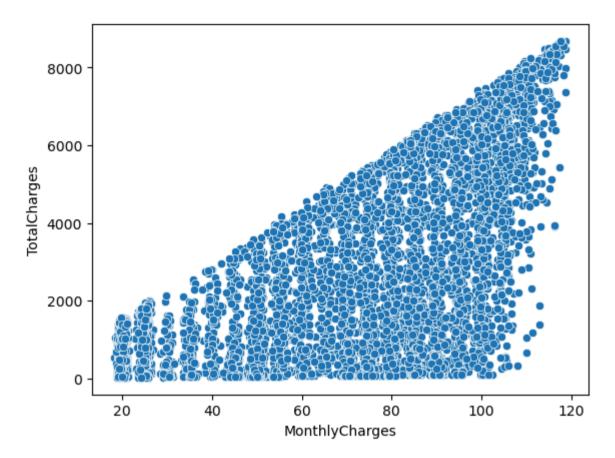
In [62]: # let's check the distribution in our dataset
sns.pairplot(data=data)

Out[62]: <seaborn.axisgrid.PairGrid at 0x220c2205850>



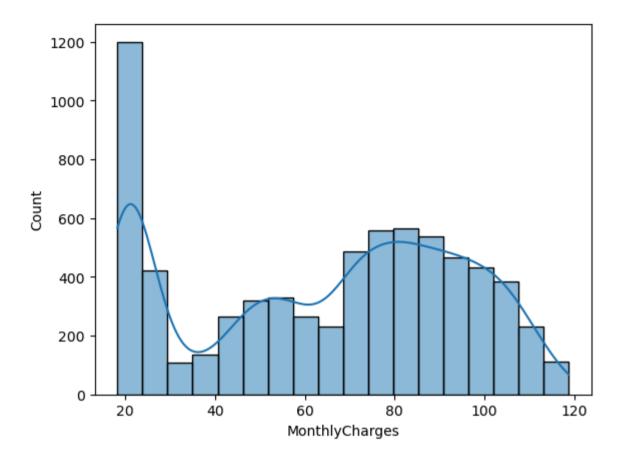


In [63]: # let's check the relation ship betweeen two variables
sns.scatterplot(x="MonthlyCharges",y="TotalCharges",data=data)
plt.show()



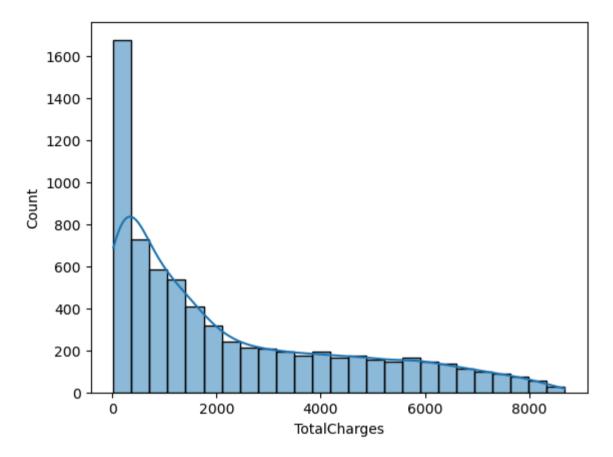
In [64]: # to check the distribution

sns.histplot(x="MonthlyCharges",data=data,kde=True)
plt.show()



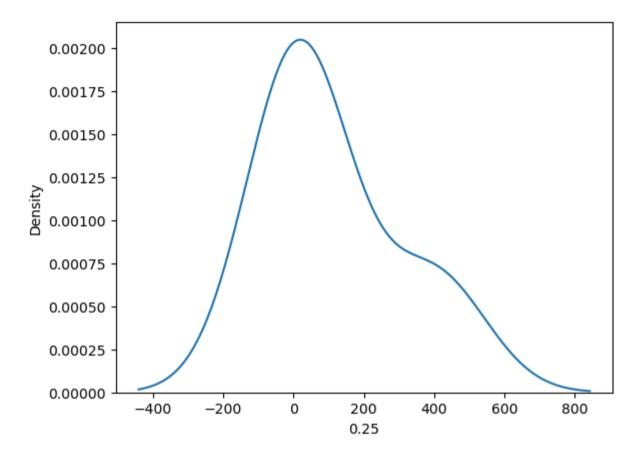
```
In [65]: # to check the distribution

sns.histplot(x="TotalCharges",data=data,kde=True)
plt.show()
```



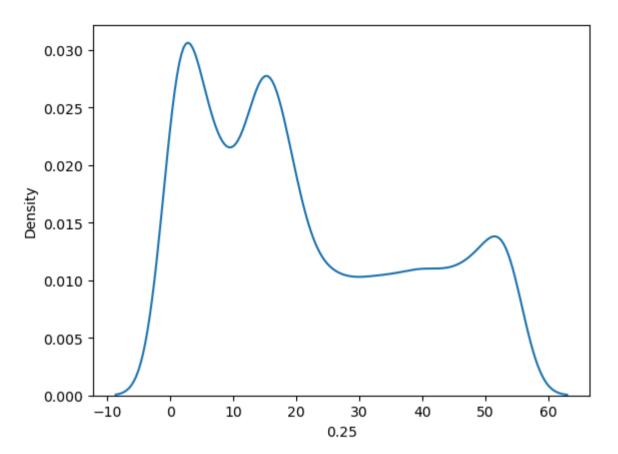
In [66]: # column wise standard deviations
data.std(axis=1,skipna=True,numeric_only=True)

```
Out[66]: 0
                    16.950147
          1
                   929.886015
          2
                    51,238804
          3
                   906.059353
          4
                    71.686777
                     . . .
          7038
                   977.768236
          7039
                  3652.505658
          7040
                   166.905998
          7041
                   144.114445
          7042
                  3393.921427
          Length: 7043, dtype: float64
In [67]: # row wise standard deviations
         data.std(axis=0,skipna=True,numeric only=True)
Out[67]: SeniorCitizen
                               0.368612
                              24.559481
          tenure
          MonthlyCharges
                              30.090047
          TotalCharges
                            2266.771362
          dtype: float64
In [68]: # column wise variance
         data.var(axis=1,skipna=True,numeric only=True)
Out[68]: 0
                  2.873075e+02
                 8.646880e+05
          1
          2
                  2.625415e+03
                  8.209436e+05
                  5.138994e+03
          7038
                  9.560307e+05
          7039
                  1.334080e+07
          7040
                  2.785761e+04
          7041
                  2.076897e+04
          7042
                  1.151870e+07
          Length: 7043, dtype: float64
```



In [71]: # Let's check the bell shape curve means proper distribution curve is made by our dataset or not? column wise
a = data.quantile(q=0.25,axis=1,numeric_only=True)
sns.kdeplot(a)

Out[71]: <Axes: xlabel='0.25', ylabel='Density'>



```
Out[73]: 0
                 7590-VHVEG
          1
                 5575-GNVDE
          2
                 3668-QPYBK
          3
                 7795-CFOCW
                 9237-H0ITU
                     . . .
          7038
                 6840-RESVB
          7039
                 2234-XADUH
          7040
                 4801-JZAZL
          7041
                 8361-LTMKD
          7042
                 3186-AJIEK
         Name: customerID, Length: 7043, dtype: category
         Categories (7043, object): ['0002-ORFBO', '0003-MKNFE', '0004-TLHLJ', '0011-IGKFF', ..., '9992-RRAMN', '9992-UJOEL', '9993-LH
         IEB', '9995-HOTOH']
In [74]: # heatmap for the correlation of the data
         sns.heatmap(data.corr(numeric only = True),cmap="Paired")
Out[74]: <Axes: >
```

