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Q.6 Write Python Programming for given Dataset "tips.csv" file.

Expected Output:

- 1. Need to use required libraries
- 2. Perform pre-processing techniques
- 3. Apply visualization modules like Box Plot, Scatter plot and explain your understanding.

```
In [1]:
```

```
#Let's start with importing required libraries
import sklearn
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

In [2]:

```
1 df = pd. read_csv ("tips.csv") # Reading the Data
2 df.head()
```

Out[2]:

	Unnamed: 0	total_bill	tip	sex	smoker	day	time	size
0	0	16.99	1.01	Female	No	Sun	Dinner	2
1	1	10.34	1.66	Male	No	Sun	Dinner	3
2	2	21.01	3.50	Male	No	Sun	Dinner	3
3	3	23.68	3.31	Male	No	Sun	Dinner	2
4	4	24.59	3.61	Female	No	Sun	Dinner	4

In [3]:

```
1 df.shape
```

Out[3]:

(244, 8)

In [4]:

```
1 df.dtypes # df.info() also this way can be done
```

Out[4]:

```
Unnamed: 0
                int64
total_bill
              float64
              float64
tip
               object
sex
smoker
               object
               object
day
time
               object
size
                int64
dtype: object
```

In [5]:

```
1 df.columns
```

Out[5]:

```
In [6]:
 1 duplicate = df[df.duplicated()]
 2 print("Duplicate Rows :")
Duplicate Rows :
In [7]:
 1 df.isnull().sum()
Out[7]:
Unnamed: 0
total_bill
              0
tip
               0
               0
sex
smoker
               0
day
               0
               0
time
size
              0
dtype: int64
In [8]:
 1 df.describe()
Out[8]:
       Unnamed: 0
                    total_bill
                                   tip
                                            size
       244.000000
                 244.000000 244.000000
                                      244 000000
 count
       121.500000
                   19.785943
                              2.998279
                                         2.569672
 mean
        70.580923
                    8.902412
                              1.383638
                                         0.951100
  std
  min
         0.000000
                    3.070000
                              1.000000
                                         1.000000
        60.750000
                   13.347500
                              2.000000
                                         2.000000
  25%
  50%
       121.500000
                   17.795000
                              2.900000
                                         2.000000
       182.250000
                  24.127500
                              3.562500
                                         3.000000
  max
       243.000000
                   50.810000
                             10.000000
                                         6.000000
In [9]:
 1 df.drop (columns = ['Unnamed: 0'],inplace=True,)
In [11]:
 1 df['day'].unique()
Out[11]:
array(['Sun', 'Sat', 'Thur', 'Fri'], dtype=object)
In [12]:
 1 df.nunique()
Out[12]:
total bill
               229
tip
               123
sex
                 2
smoker
                 2
day
                 4
time
size
                 6
dtype: int64
In [13]:
 1 df.dtypes
Out[13]:
total_bill
               float64
tip
               float64
                object
sex
smoker
                object
day
                object
time
                object
size
                 int64
dtype: object
```

univariate analysis of continuous variables

total_bill

```
In [14]:
 1 df['total_bill'].describe()
Out[14]:
         244.000000
19.785943
count
mean
           8.902412
std
min
           3.070000
          13.347500
50%
          17.795000
75%
          24.127500
          50.810000
max
Name: total_bill, dtype: float64
In [15]:
 1 df['total_bill'].skew()
Out[15]:
1.1332130376158205
In [16]:
 1 sns.distplot(df['total_bill'])
Out[16]:
<AxesSubplot:xlabel='total_bill', ylabel='Density'>
   0.06
   0.05
   0.04
   0.03
   0.02
   0.01
   0.00
                        20
In [18]:
 1 sns.boxplot(df['total_bill'])
Out[18]:
<AxesSubplot:xlabel='total_bill'>
```

total_bill

50

tip

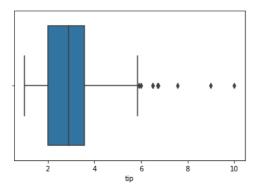
```
In [19]:
 1 df['tip'].describe()
Out[19]:
         244.000000
count
           2.998279
mean
           1.383638
std
min
           1.000000
25%
           2.000000
50%
           2.900000
           3.562500
          10.000000
max
Name: tip, dtype: float64
In [20]:
 1 df['tip'].skew()
Out[20]:
1.4654510370979401
In [21]:
 1 sns.distplot(df['tip'])
<AxesSubplot:xlabel='tip', ylabel='Density'>
   0.4
  0.3
Density
0.0
   0.1
   0.0
                                           10
```

In [22]:

```
1 sns.boxplot(df['tip'])
```

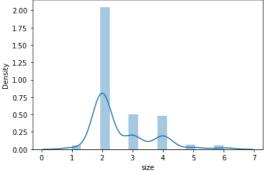
Out[22]:

<AxesSubplot:xlabel='tip'>



size

```
In [23]:
 1 print('unique categories--->',df['size'].unique())
 2 #counting the uniques
 3 print('value_counts for each unique categories---->',df['size'].value_counts())
 4 print('Null value--->',df['size'].isnull().sum())
unique categories---> [2 3 4 1 6 5]
value_counts for each unique categories----> 2
3
      38
4
      37
       4
1
6
Name: size, dtype: int64
Null value---> 0
In [24]:
 1 df['size'].describe()
Out[24]:
count
         244.000000
           2.569672
mean
           0.951100
std
           1.000000
min
           2.000000
25%
           2.000000
50%
75%
           3.000000
           6.000000
max
Name: size, dtype: float64
In [25]:
 1 df['size'].skew()
Out[25]:
1.4478815386834785
In [27]:
 1 sns.distplot(df['size'])
Out[27]:
<AxesSubplot:xlabel='size', ylabel='Density'>
   2.00
  1.75
  1.50
```

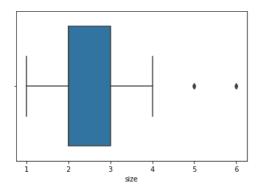


```
In [26]:
```

```
1 sns.boxplot(df['size'])
```

Out[26]:

<AxesSubplot:xlabel='size'>

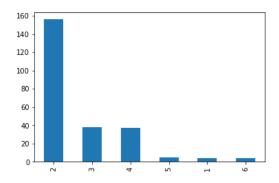


In [28]:

```
1 df['size'].value_counts().plot(kind='bar')
```

Out[28]:

<AxesSubplot:>



sex

```
In [29]:
```

```
print('unique categories--->',df['sex'].unique())

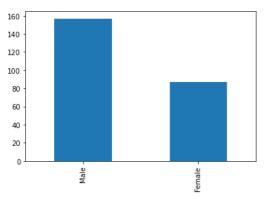
#counting the uniques
print('value_counts for each unique categories---->',df['sex'].value_counts())
print('Null value--->',df['sex'].isnull().sum())
```

In [30]:

```
1 df['sex'].value_counts().plot(kind='bar')
```

Out[30]:

<AxesSubplot:>



smoker

```
In [31]:
 1 print('unique categories--->',df['smoker'].unique())
 2 #counting the uniques
 3 print('value_counts for each unique categories---->',df['smoker'].value_counts())
 4 print('Null value--->',df['smoker'].isnull().sum())
unique categories---> ['No' 'Yes']
value_counts for each unique categories----> No
                                                     151
Yes
        93
Name: smoker, dtype: int64
Null value---> 0
In [32]:
 1 df['smoker'].value_counts().plot(kind='bar')
Out[32]:
<AxesSubplot:>
 140
 120
 100
  80
  60
  40
  20
  0
```

day

```
In [33]:
```

```
print('unique categories--->',df['day'].unique())

#counting the uniques
print('value_counts for each unique categories---->',df['day'].value_counts())

print('Null value--->',df['day'].isnull().sum())

unique categories---> ['Sun' 'Sat' 'Thur' 'Fri']
value_counts for each unique categories----> Sat 87
Sun 76
Thure 62
```

Thur 62 Fri 19

Name: day, dtype: int64

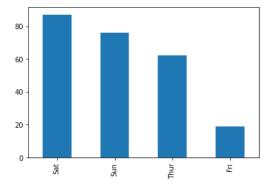
Null value---> 0

In [34]:

```
1 df['day'].value_counts().plot(kind='bar')
```

Out[34]:

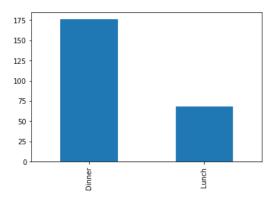
<AxesSubplot:>



time

```
In [36]:
    print('unique categories--->',df['time'].unique())
 2 #counting the uniques
 3 print('value_counts for each unique categories---->',df['time'].value_counts())
 4 print('Null value--->',df['time'].isnull().sum())
unique categories---> ['Dinner' 'Lunch']
value_counts for each unique categories----> Dinner
                                                       176
Lunch
Name: time, dtype: int64
Null value---> 0
In [37]:
 1 df['time'].value_counts().plot(kind='bar')
Out[37]:
```

<AxesSubplot:>



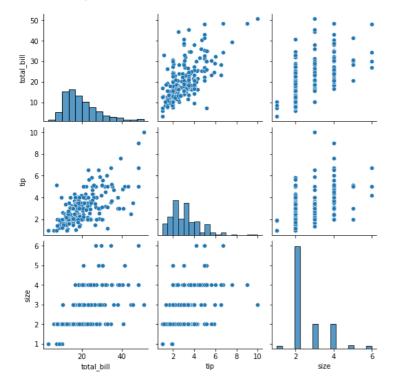
multivariate analysis

```
In [38]:
```

```
1 sns.pairplot(df)
```

Out[38]:

<seaborn.axisgrid.PairGrid at 0x1b3dfdd8a30>



```
In [ ]:
```

1

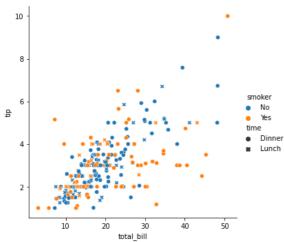
Type $\it Markdown$ and LaTeX: $\it \alpha^2$

Type $\mathit{Markdown}$ and LaTeX : α^2

Type *Markdown* and LaTeX: α^2

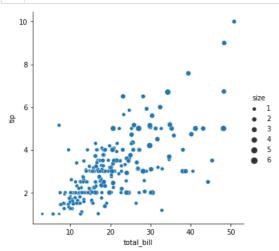
```
In [41]:
```

```
sns.relplot (x = 'total_bill', y = 'tip', data = df, hue = 'smoker', style ='time')
plt.show()
```



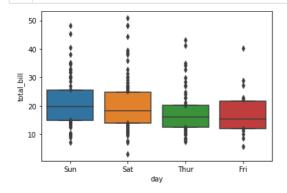
In [43]:

```
1 sns.relplot(x = 'total_bill', y = 'tip', data= df, size = 'size')
2 plt.show()
```



In [45]:

```
1 sns.boxplot(x='day',y='total_bill',data=df,whis=False)
2 plt.show()
```



In [47]:

```
1 x=pd.DataFrame(pd.pivot_table(df,index=['sex','smoker'],aggfunc='count')['total_bill'])
```

```
In [48]:
```

```
1 x
```

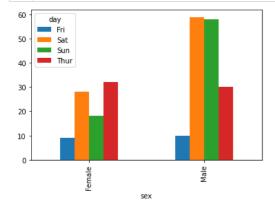
Out[48]:

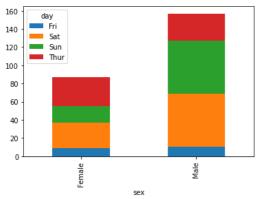
total_bill	
------------	--

sex	smoker	
Female	No	54
remale	Yes	33
Male	No	97
Wate	Yes	60

In [50]:

```
pd.crosstab(df['sex'],df['day']).plot(kind='bar')
plt.show()
pd.crosstab(df['sex'],df['day']).plot(kind='bar',stacked=True)
plt.show()
```



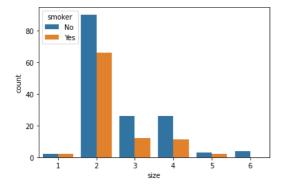


In [53]:

```
1 sns.countplot(
2 x='size', hue='smoker',data=df)
```

Out[53]:

<AxesSubplot:xlabel='size', ylabel='count'>



Maximum pepole are coming in size of 2 then followed by 3, 4

maximum people prefer to go outside on Saturday and Sunday

Those people who are visiting single have an equal chance to smoke after lunch or dinner

people mostly go outise for the dinner

Based on the scatter plot, it's possible to see if there is a positive correlation between the total bill and the tip and size amount, meaning that as the total bill increases, so does the tip amount, which could suggest that customers tend to tip a higher percentage of the total bill for higher bill amounts. Or it could also indicate that when customers spend more, they tend to tip more.

Based on the scatter plot, it's possible to see if there is a positive correlation between the total bill and the party size, meaning that as the size of the party increases, so does the total bill, which could suggest that customers tend to spend more when there are more people in their party.

Based on the scatter plot, it's possible to see if there is a positive correlation between the tip amount and the party size, meaning that as the size of the party increases, so does the tip amount, which could suggest that customers tend to tip more when there are more people in their party.

In []: