

5- Seaborn

November 25, 2025

0.1 Seaborn

0.2 Distribution plots

- distplot
- joinplot
- pairplot

```
[9]: import seaborn as sns
```

```
[11]: df=sns.load_dataset("tips")
```

```
[4]: df.head()
```

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

0.2.1 Correlation with Heatmap

A correlation heatmap uses colored cells, typically in a monochromatic scale, to show a 2D correlation matrix (table) between two discrete dimensions or event types. It is very important in Feature Selection

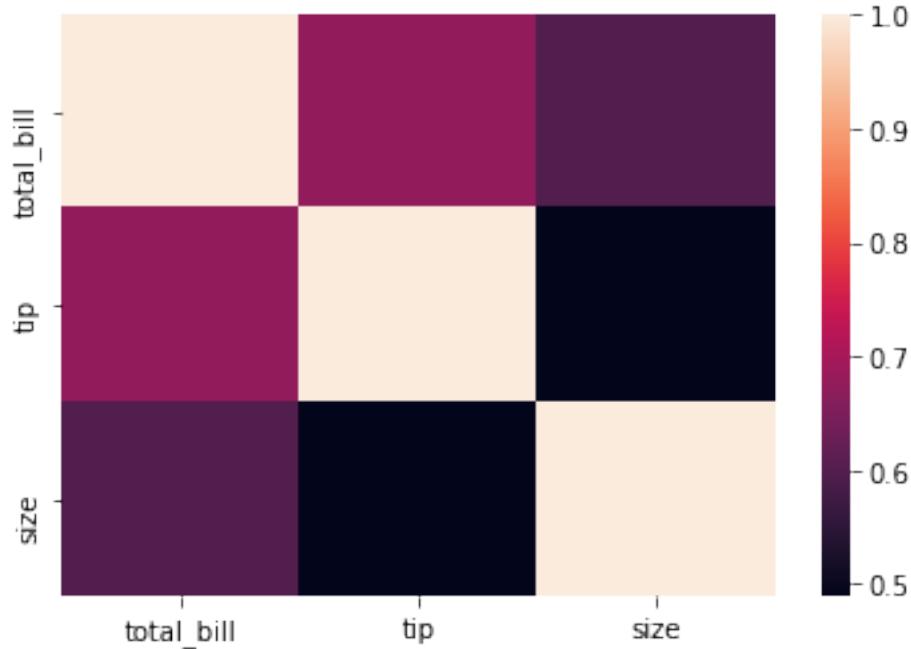
```
[45]: df.corr()
```

```
[45]:      total_bill      tip      size
total_bill    1.000000  0.675734  0.598315
tip          0.675734  1.000000  0.489299
size         0.598315  0.489299  1.000000
```

It calculates how strongly numerical columns in your DataFrame are related to each other. *
Output: a table (matrix) * Values range from -1 to +1

```
[47]: sns.heatmap(df.corr())
```

```
[47]: <matplotlib.axes._subplots.AxesSubplot at 0x26d290442e8>
```

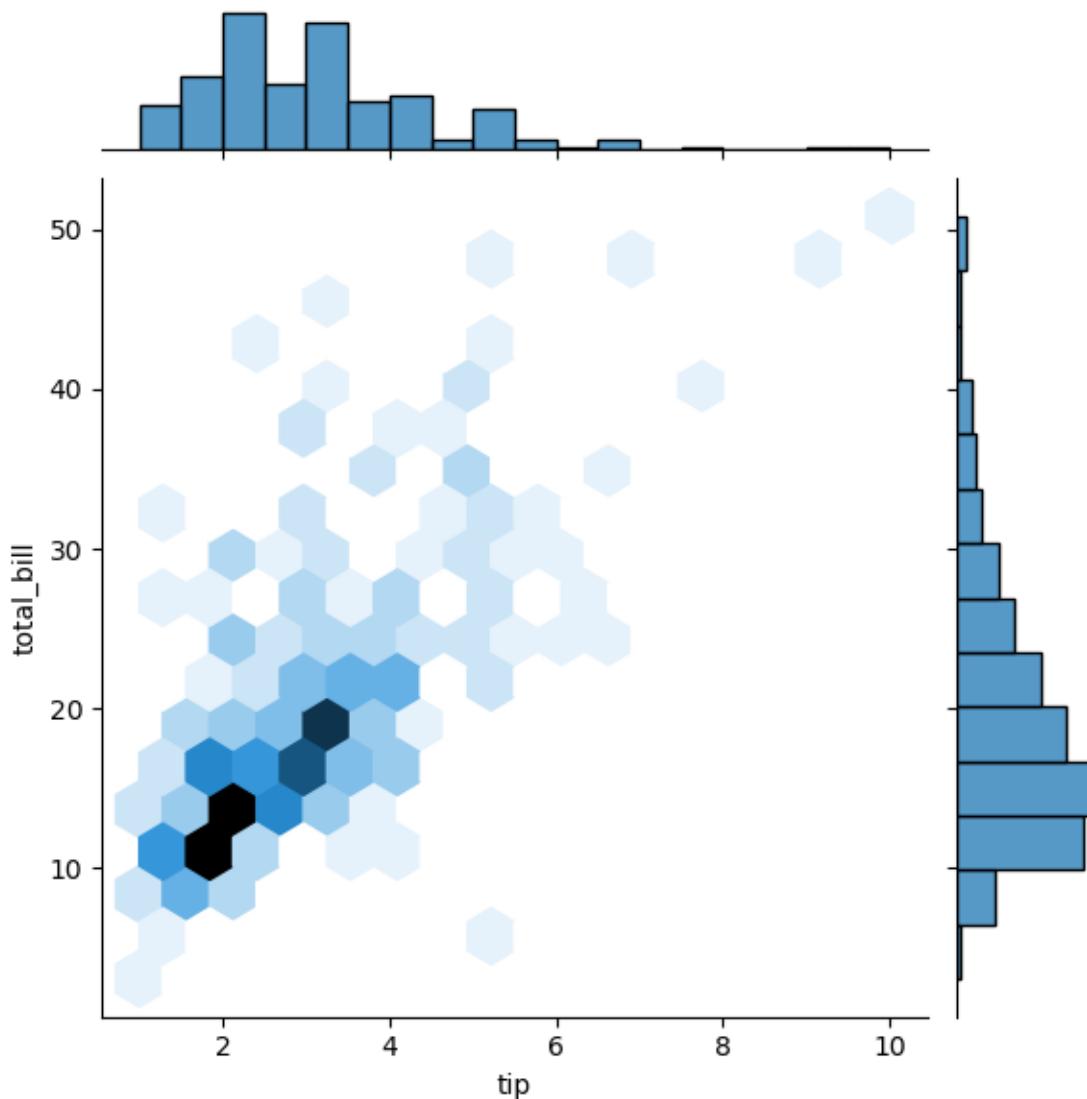


0.3 JoinPlot / scatterplot

A join plot allows to study the relationship between 2 numeric variables. The central chart display their correlation. It is usually a scatterplot, a hexbin plot, a 2D histogram or a 2D density plot
Univariate Analysis

```
[21]: sns.jointplot(x='tip',y='total_bill',data=df,kind='hex') #Other options for kind:  
      ↳ 'scatter', 'kde', 'reg', 'resid'.
```

```
[21]: <seaborn.axisgrid.JointGrid at 0x16ab84c80>
```

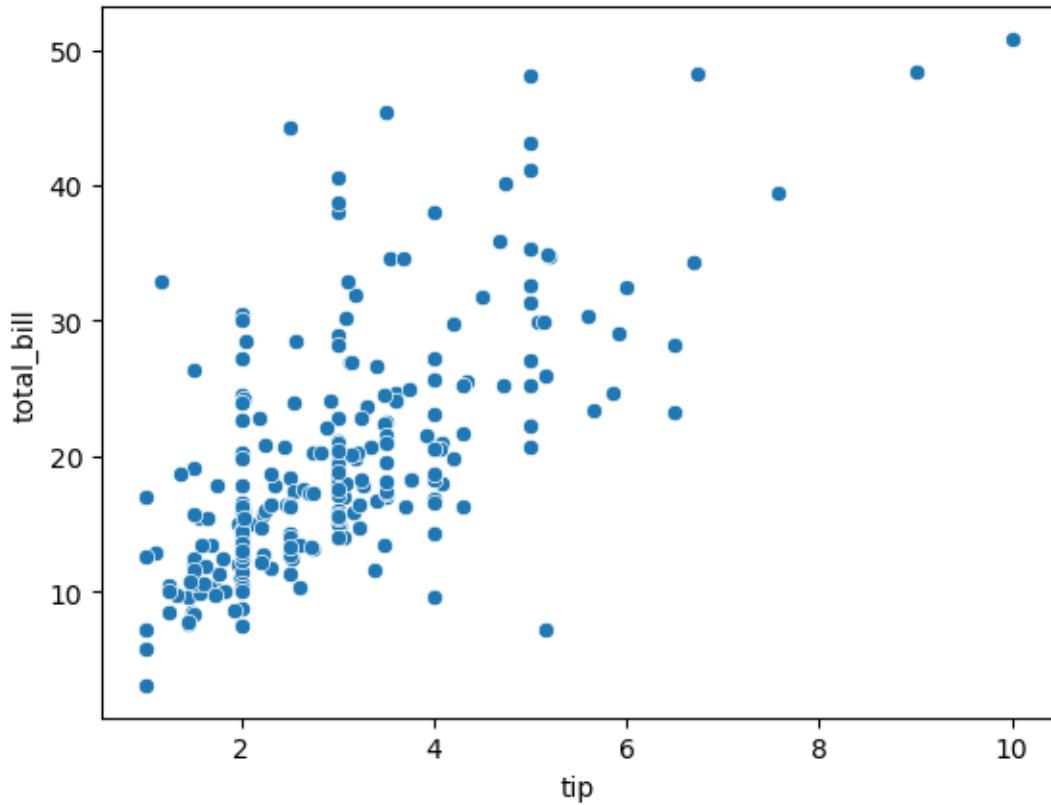


0.4 Features of a scatter plot:

- Each point represents one row in the dataset
- Shows the relationship between tip and total_bill

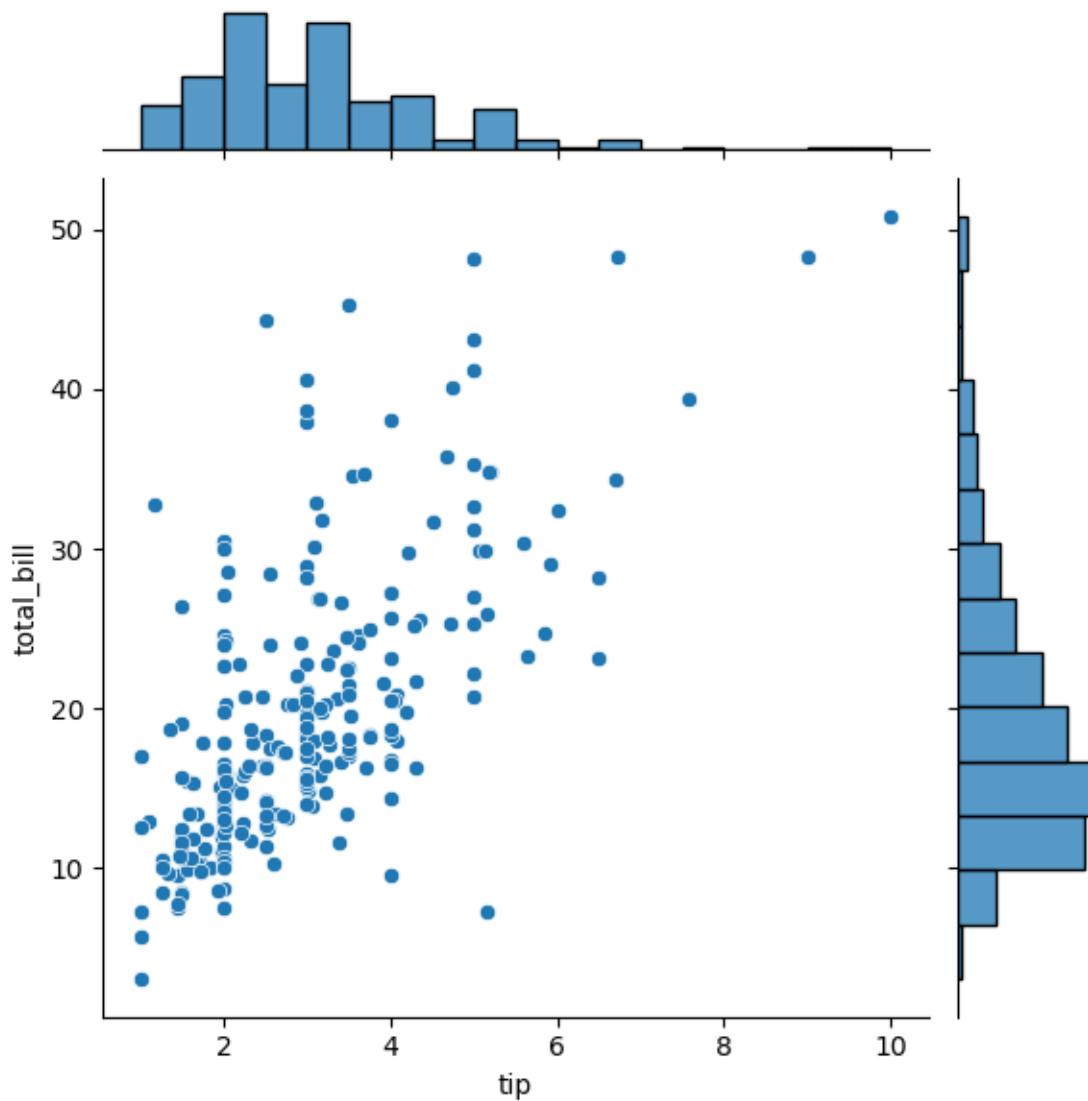
```
[17]: # Scatter plot
sns.scatterplot(x='tip', y='total_bill', data=df)
```

```
[17]: <Axes: xlabel='tip', ylabel='total_bill'>
```



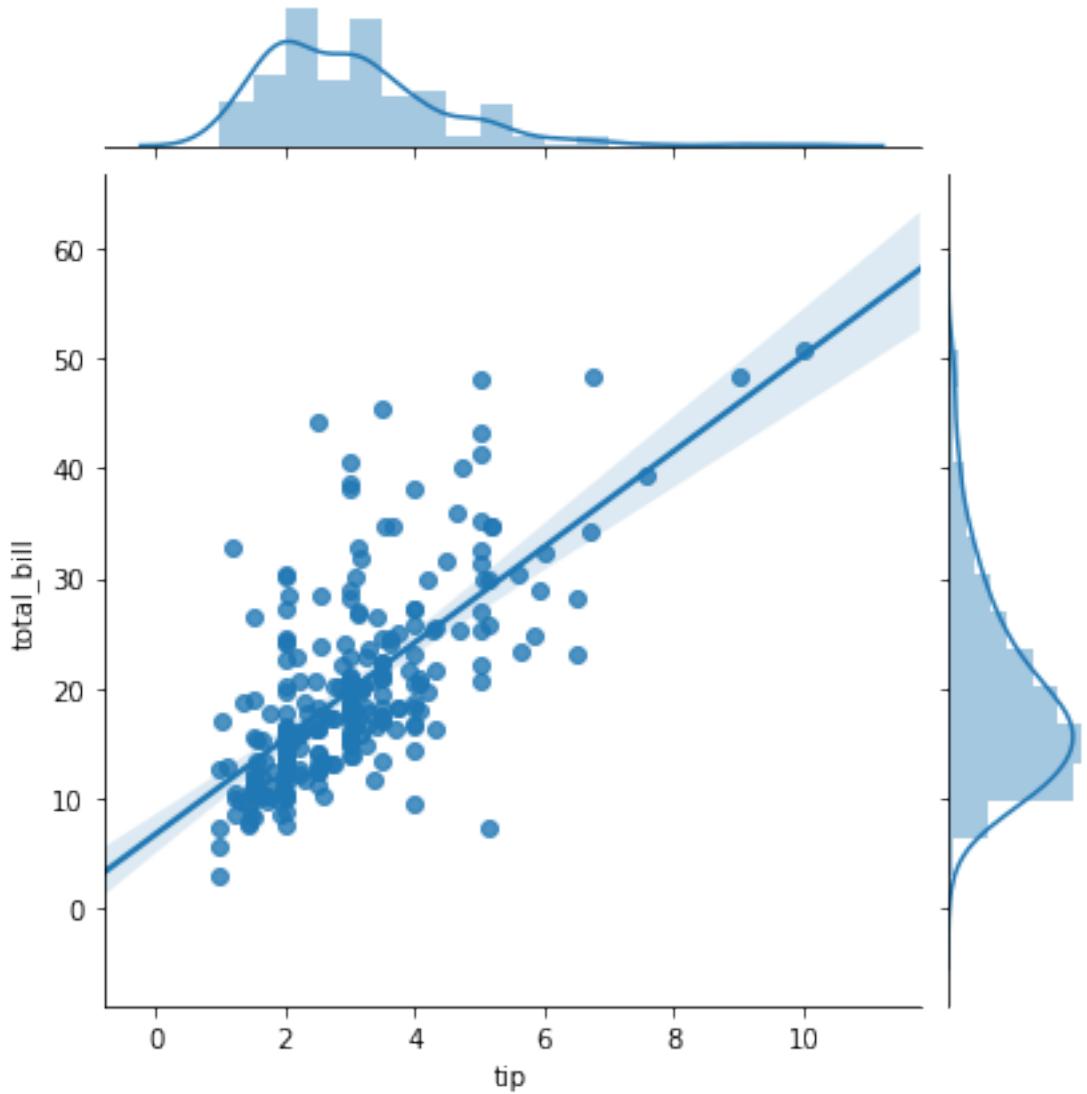
```
[19]: sns.jointplot(x='tip', y='total_bill', data=df, kind='scatter')
```

```
[19]: <seaborn.axisgrid.JointGrid at 0x16a98fe60>
```



```
[21]: sns.jointplot(x='tip',y='total_bill',data=df,kind='reg')
```

```
[21]: <seaborn.axisgrid.JointGrid at 0x26d268a2780>
```

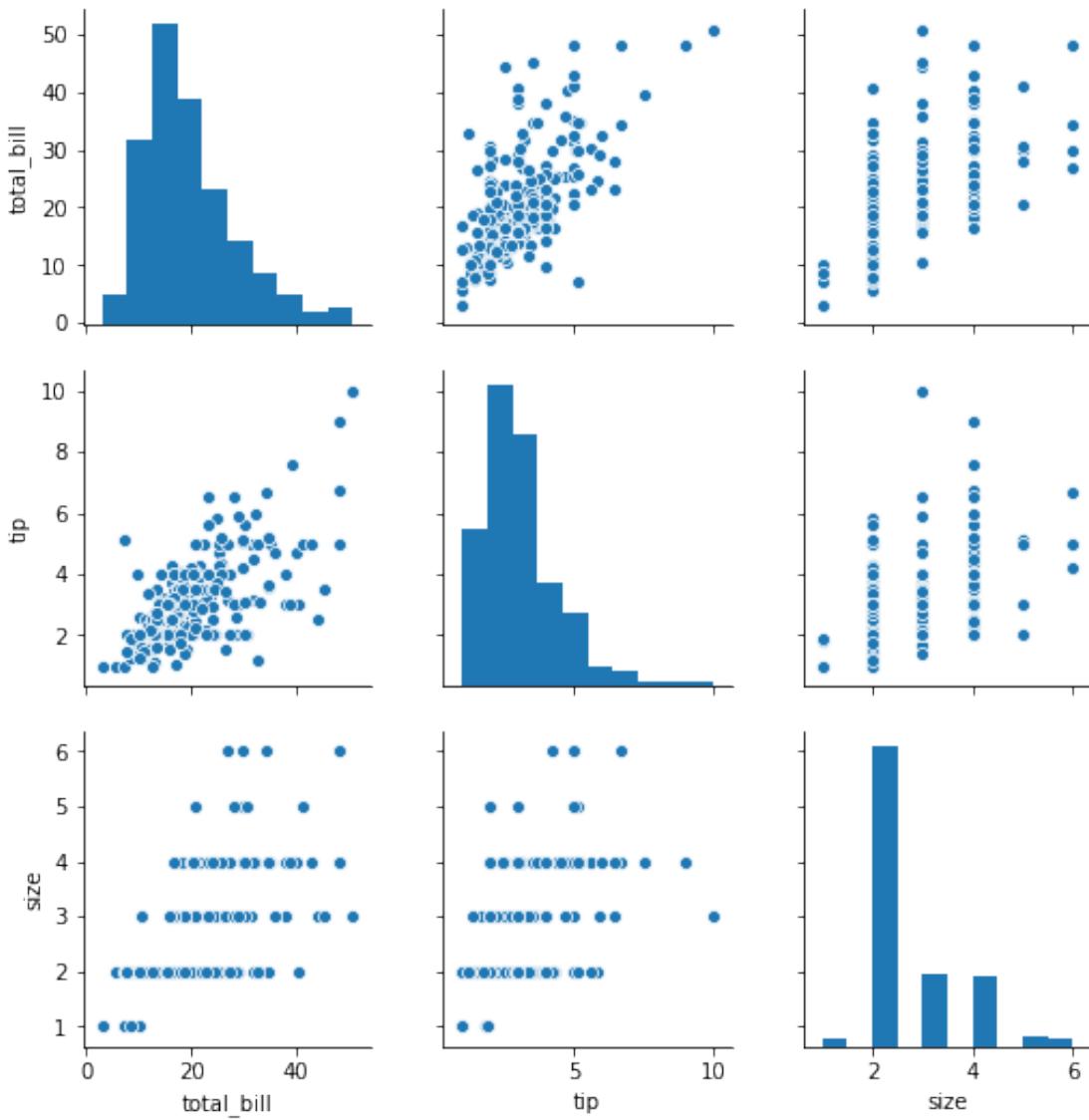


0.5 Pair plot

A “pairs plot” is also known as a scatterplot, in which one variable in the same data row is matched with another variable’s value, like this: Pairs plots are just elaborations on this, showing all variables paired with all the other variables

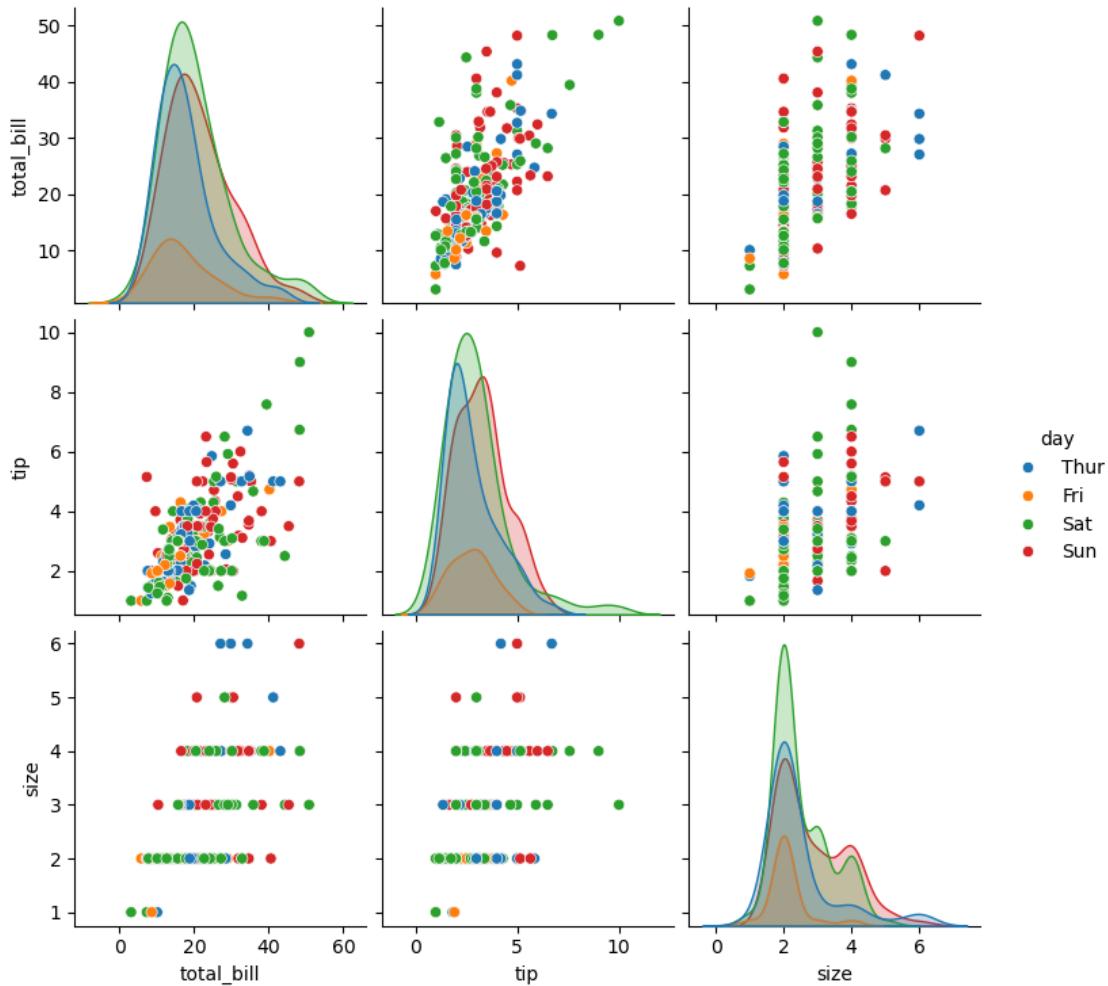
```
[8]: sns.pairplot(df)
```

```
[8]: <seaborn.axisgrid.PairGrid at 0x26d24010630>
```



```
[26]: sns.pairplot(df,hue='day')
```

```
[26]: <seaborn.axisgrid.PairGrid at 0x16ab84a10>
```



0.6 Dist plot

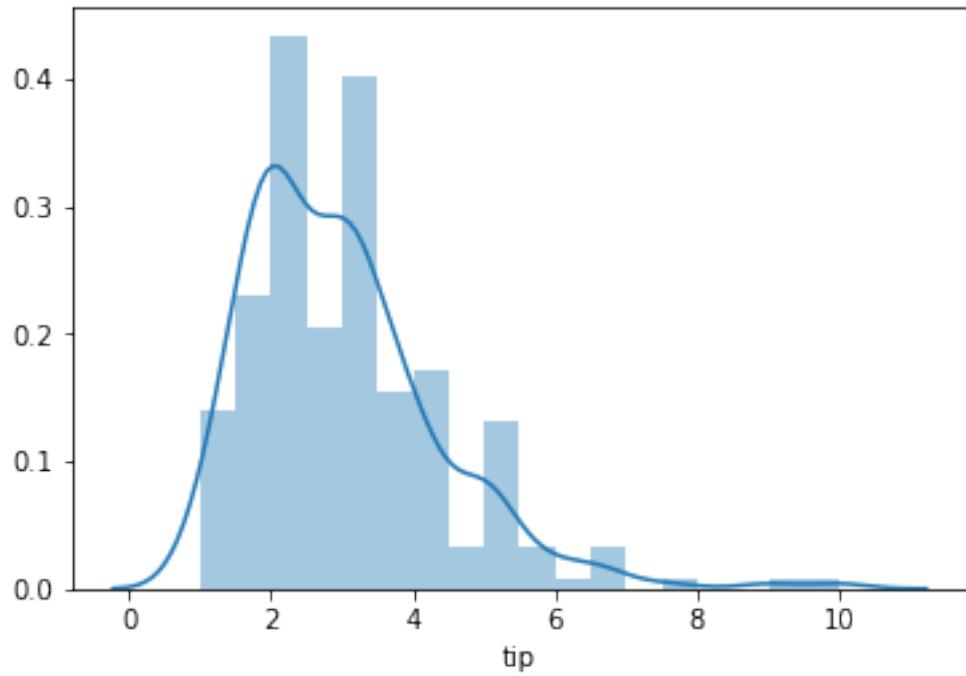
Dist plot helps us to check the distribution of the columns feature

```
[11]: sns.distplot(df['tip'])
```

```
C:\Users\krish.naik\AppData\Local\Continuum\anaconda3\envs\myenv\lib\site-
packages\scipy\stats\stats.py:1713: FutureWarning: Using a non-tuple sequence
for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of
`arr[seq]`. In the future this will be interpreted as an array index,
`arr[np.array(seq)]`, which will result either in an error or a different
result.

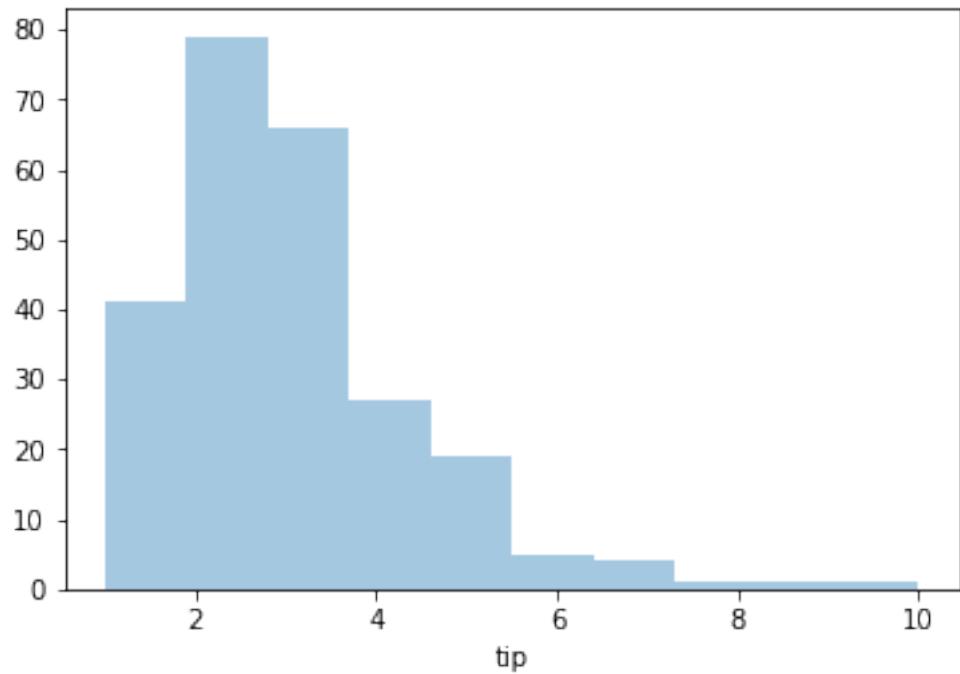
    return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval
```

```
[11]: <matplotlib.axes._subplots.AxesSubplot at 0x26d24a32be0>
```



```
[13]: sns.distplot(df['tip'], kde=False, bins=10)
```

```
[13]: <matplotlib.axes._subplots.AxesSubplot at 0x26d24aec668>
```



0.7 Categorical Plots

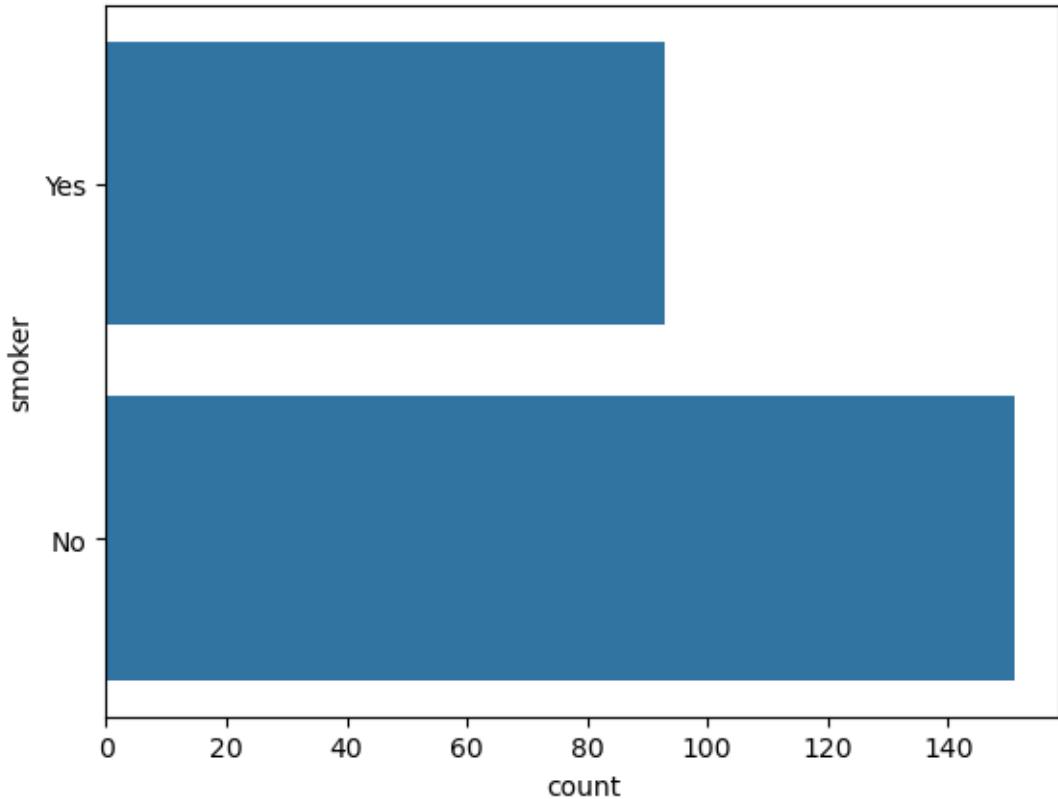
Seaborn also helps us in doing the analysis on Categorical Data points.

- boxplot
- violinplot
- countplot
- bar plot

```
[38]: ## Count plot
```

```
sns.countplot(y='smoker', data=df)
```

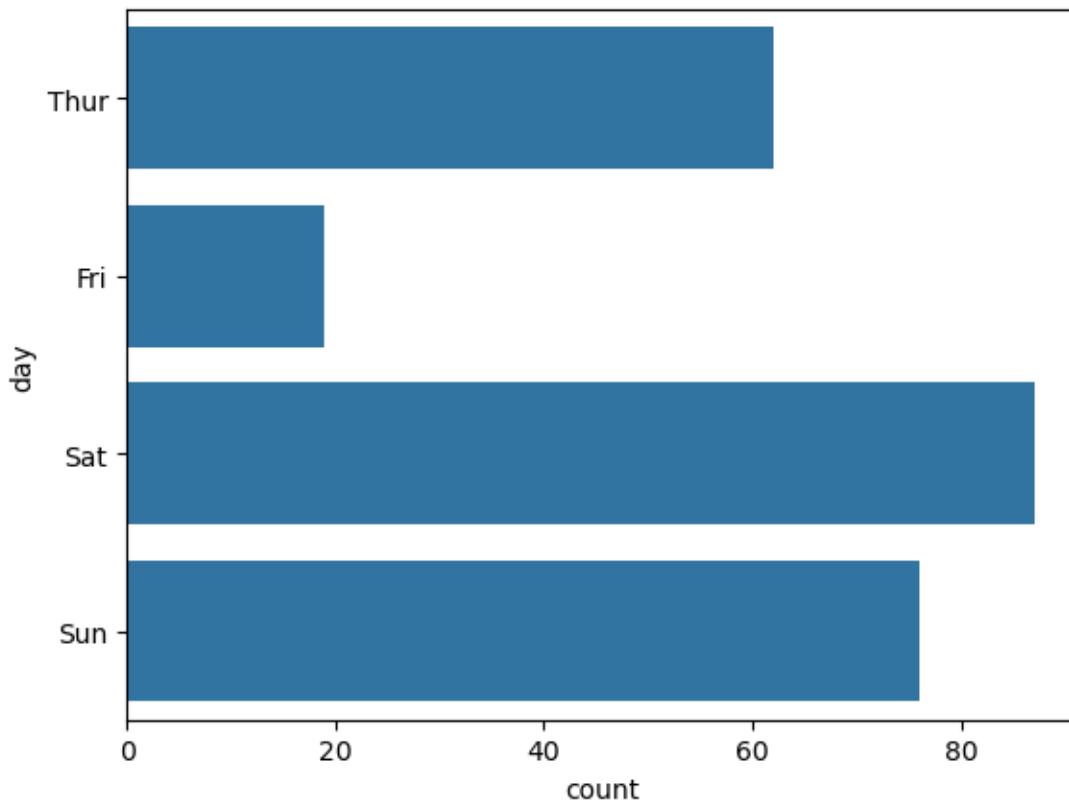
```
[38]: <Axes: xlabel='count', ylabel='smoker'>
```



```
[34]: ## Count plot
```

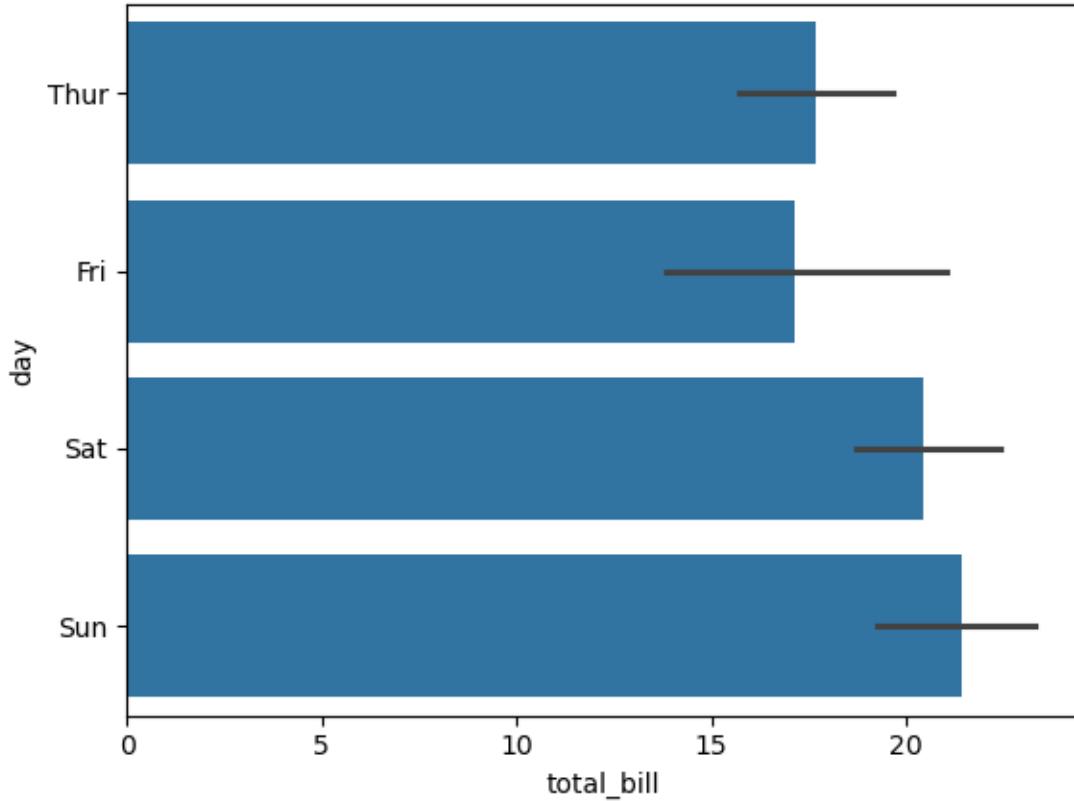
```
sns.countplot(y='day', data=df)
```

```
[34]: <Axes: xlabel='count', ylabel='day'>
```



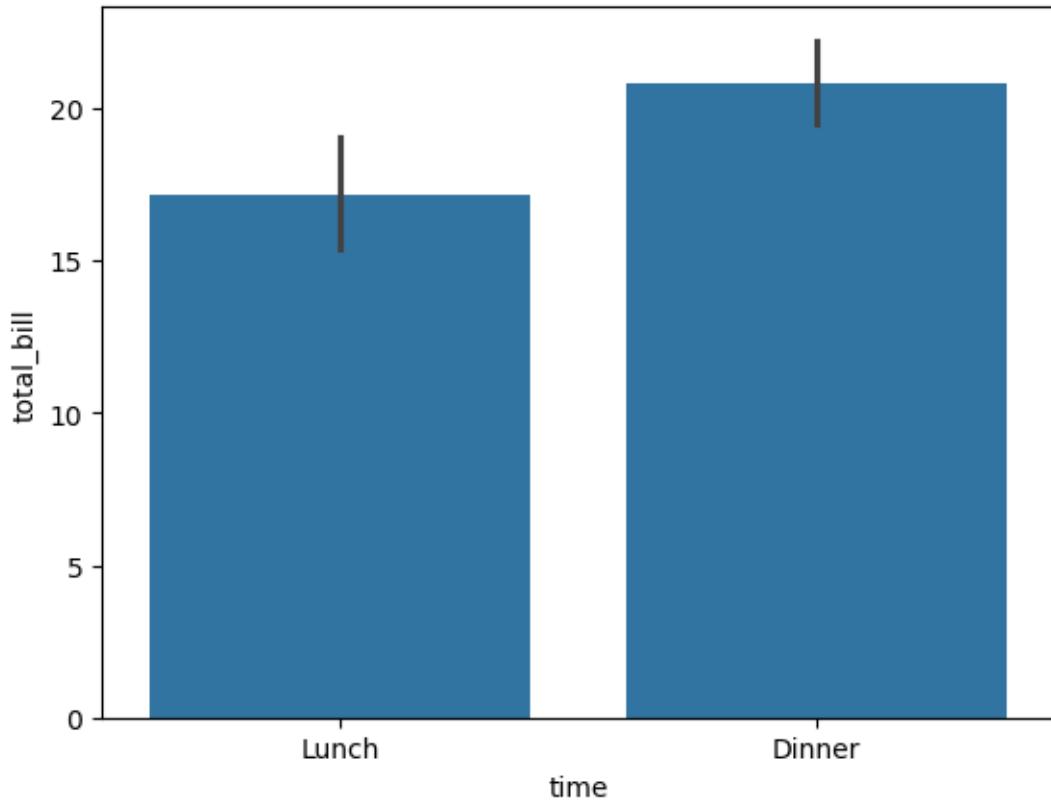
```
[40]: ## Bar plot  
sns.barplot(x='total_bill',y='day',data=df)
```

```
[40]: <Axes: xlabel='total_bill', ylabel='day'>
```



```
[44]: ## Bar plot
sns.barplot(x='time',y='total_bill',data=df)
```

```
[44]: <Axes: xlabel='time', ylabel='total_bill'>
```



```
[30]: df.head()
```

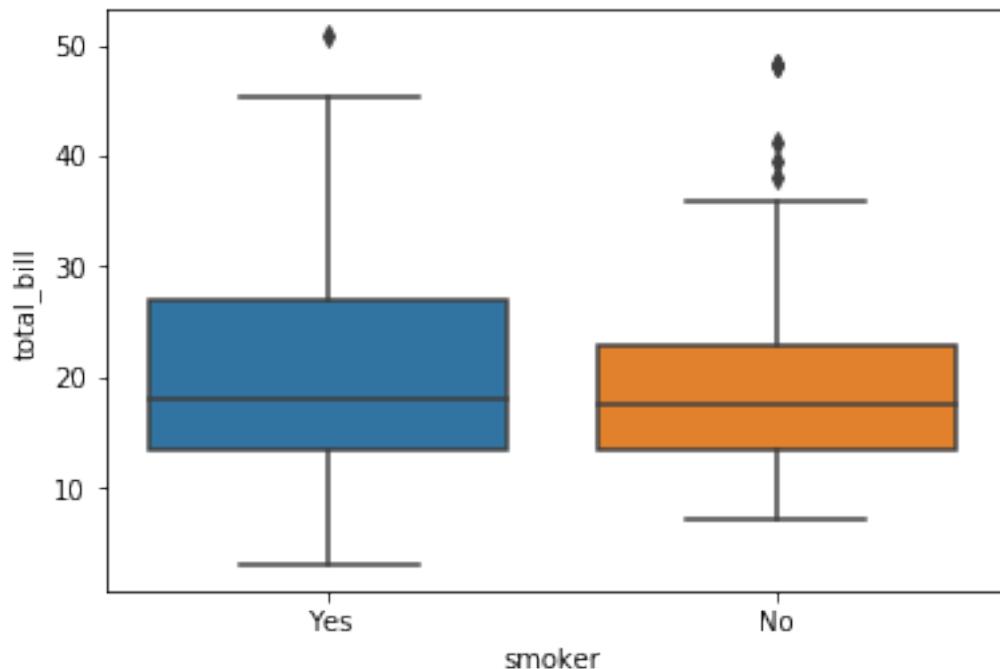
```
[30]:    total_bill  tip      sex smoker  day   time  size
 0       16.99  1.01  Female     No  Sun  Dinner     2
 1       10.34  1.66    Male     No  Sun  Dinner     3
 2       21.01  3.50    Male     No  Sun  Dinner     3
 3       23.68  3.31    Male     No  Sun  Dinner     2
 4       24.59  3.61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

0.8 Box plot

A box and whisker plot (sometimes called a boxplot) is a graph that presents information from a five-number summary.

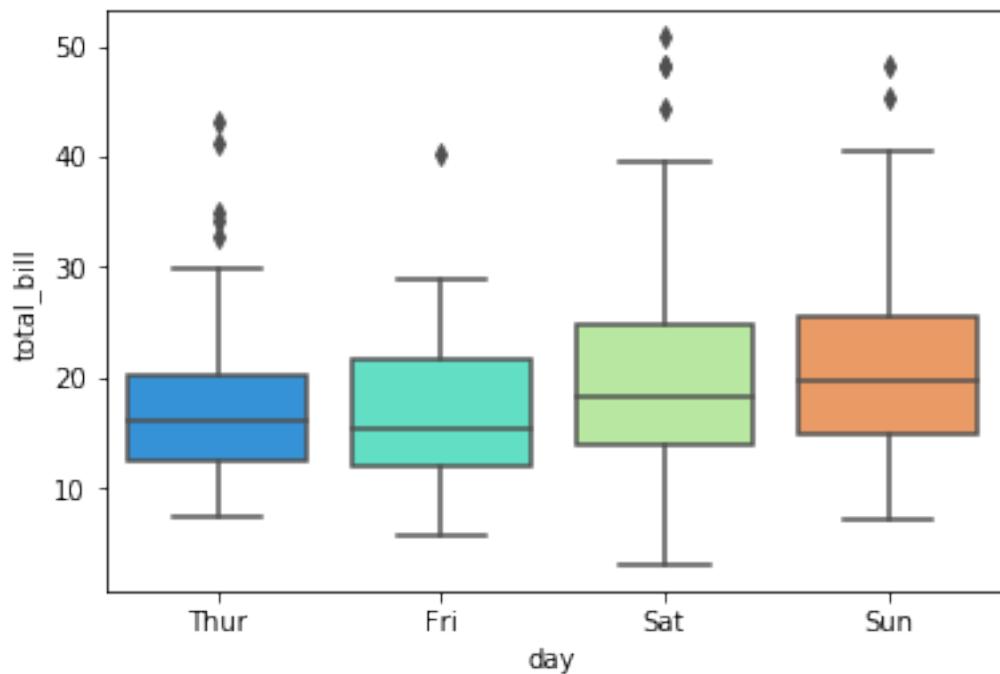
```
[31]: sns.boxplot('smoker','total_bill', data=df)
```

```
[31]: <matplotlib.axes._subplots.AxesSubplot at 0x26d287f7d30>
```



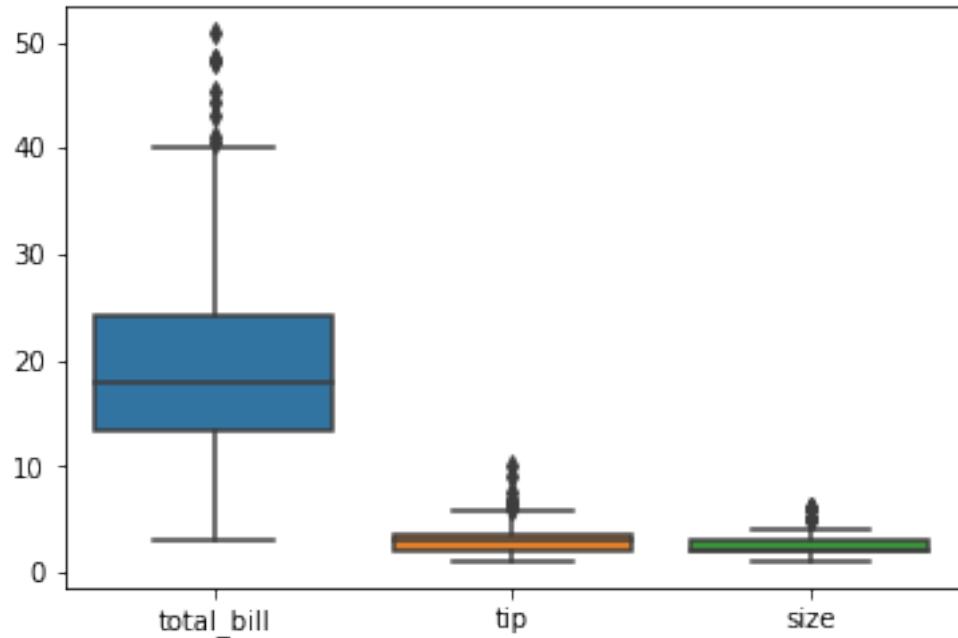
```
[33]: sns.boxplot(x="day", y="total_bill", data=df, palette='rainbow')
```

```
[33]: <matplotlib.axes._subplots.AxesSubplot at 0x26d2885f940>
```



```
[37]: sns.boxplot(data=df,orient='v')
```

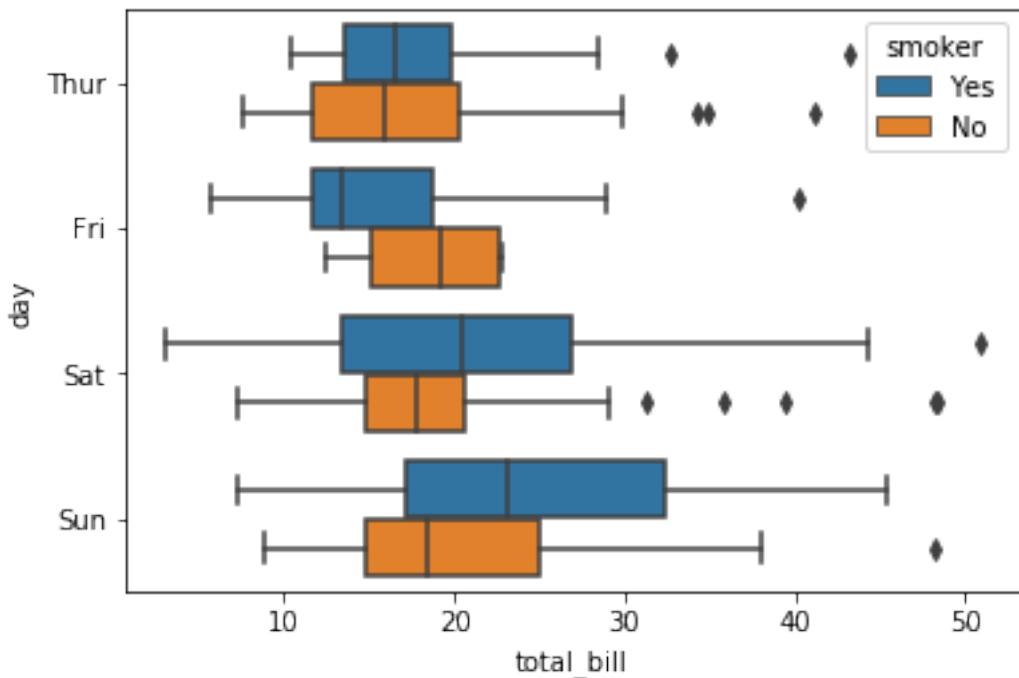
```
[37]: <matplotlib.axes._subplots.AxesSubplot at 0x26d28b23748>
```



```
[39]: # categorize my data based on some other categories
```

```
sns.boxplot(x="total_bill", y="day", hue="smoker", data=df)
```

```
[39]: <matplotlib.axes._subplots.AxesSubplot at 0x26d28c6ecf8>
```

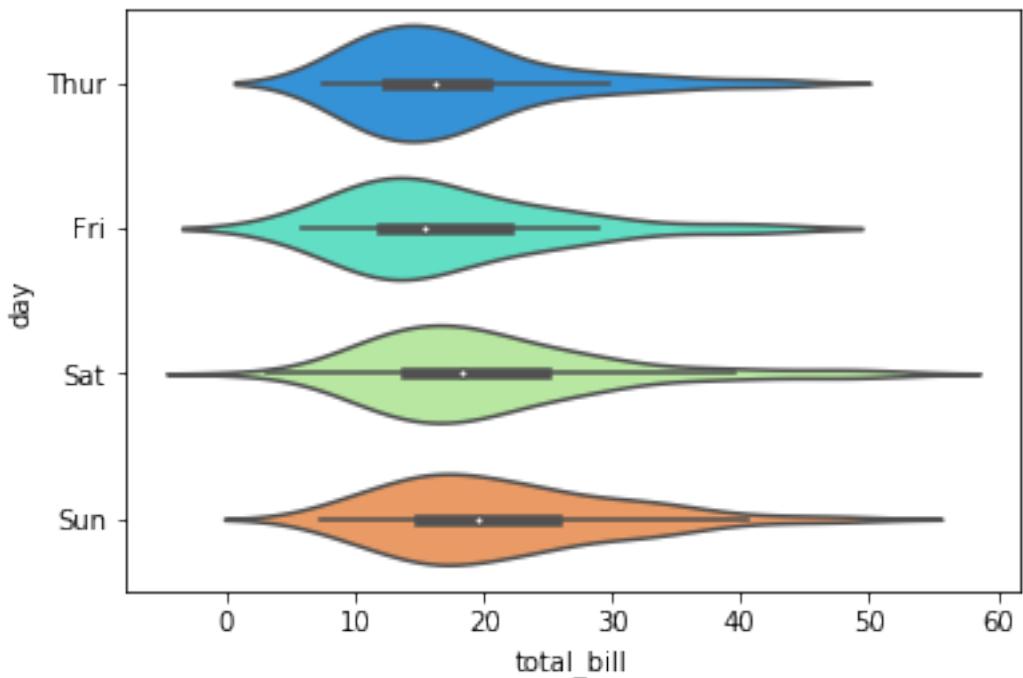


0.9 Violin Plot

Violin plot helps us to see both the distribution of data in terms of Kernel density estimation and the box plot

```
[41]: sns.violinplot(x="total_bill", y="day", data=df, palette='rainbow')
```

```
[41]: <matplotlib.axes._subplots.AxesSubplot at 0x26d28d5ca20>
```



```
[ ]: ## Practise Homework: use iris dataset and plot all graphs
```

```
iris = sns.load_dataset('iris')
```

```
[ ]:
```