

```
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
iris = sns.load_dataset("iris")
```

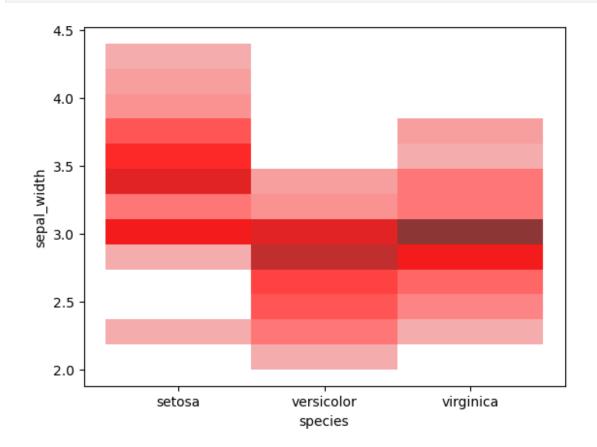
#### **Distribution Plots**

- Distribution Plots are used for examining univariate and bivariate distributions meaning such distributions that involve one variable or two discrete variables.
- There are various types of categorical plots:
  - a. Histogram
  - b. Distplot
  - c. Pairplot
  - d. KDE Plot

### 1. Histogram:

- A histogram is basically used to represent data provided in a form of some groups.
- It is accurate method for the graphical representation of numerical data distribution.
- It can be plotted using the histplot() function.
- Syntax:

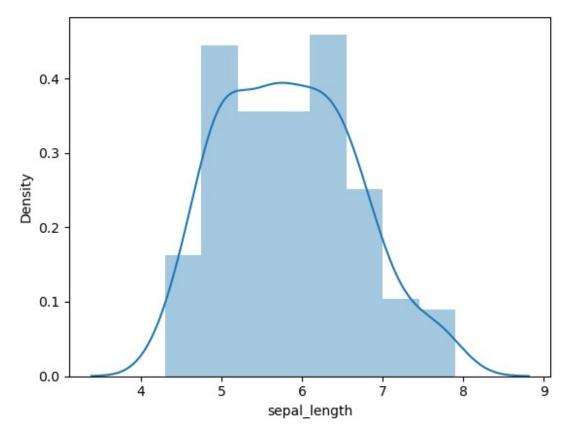
- histplot(data=None, \*, x=None, y=None, hue=None, \*\*kwargs)
sns.histplot(x='species', y='sepal\_width', data=iris, color='red')
plt.show()



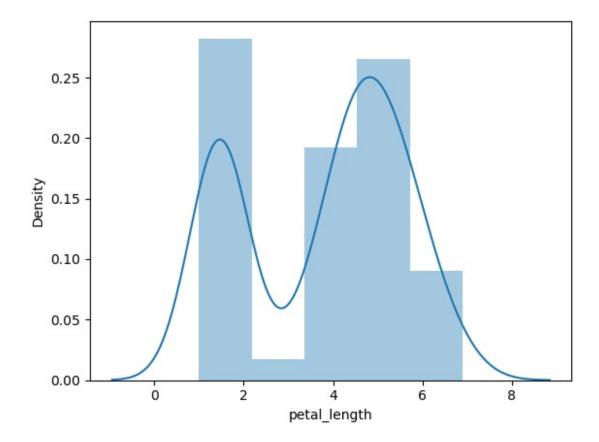
### 2. Distplot:

- Distplot is used basically for univariant set of observations and visualizes it through a histogram i.e. only one observation and hence we choose one particular column of the dataset.
- It is potted using the distplot() method.
- Syntax:
  - distplot(a[, bins, hist, kde, rug, fit, ...])

```
sns.distplot(iris['sepal_length'])
plt.show()
```



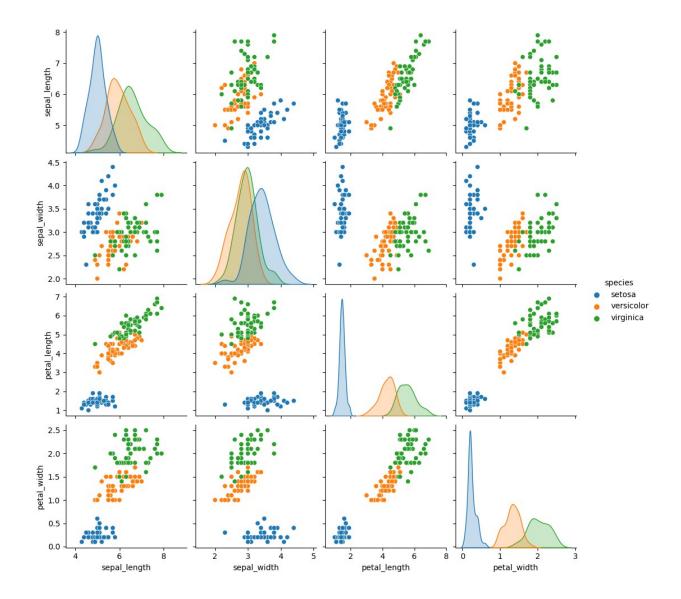
```
sns.distplot(iris['petal_length'])
plt.show()
```



## 3. Pairplot:

- Pairplot represents pairwise relation across the entire dataframe and supports an additional argument called hue for categorical separation.
- It is plotted using the pairplot() method.
- Syntax:
  - pairplot(data[, hue, hue\_order, palette, ...])

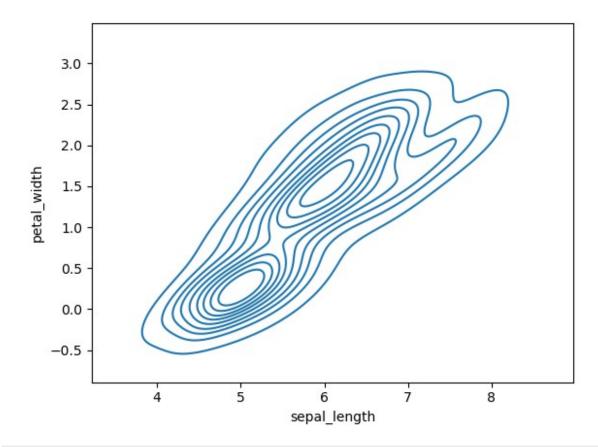
```
sns.pairplot(data=iris, hue='species')
<seaborn.axisgrid.PairGrid at 0x1f8b8006510>
```



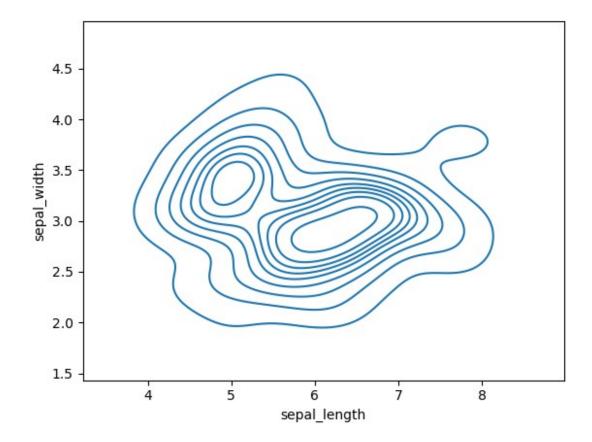
#### 4. KDE Plot:

- KDE Plot described as Kernel Density Estimate is used for visualizing the Probability Density of a continuous variable.
- It depicts the probability density at different values in a continuous variable.
- We can also plot a single graph for multiple samples which helps in more efficient data visualization.
- Syntax:
  - seaborn.kdeplot(x=None, \*, y=None, vertical=False, palette=None, \*\*kwargs)

```
sns.kdeplot(x='sepal_length', y='petal_width', data=iris)
plt.show()
```



sns.kdeplot(x='sepal\_length', y='sepal\_width', data=data)
plt.show()



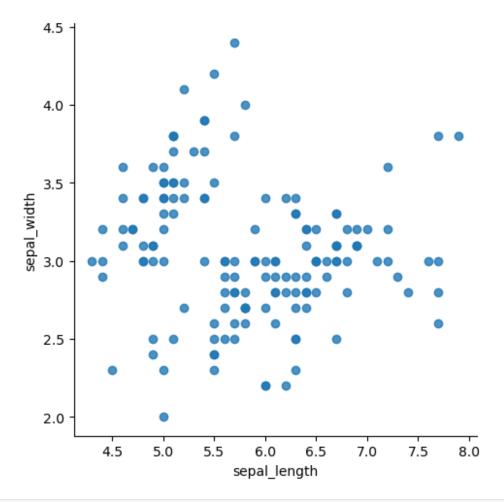
# **Regression Plots**

- Regression plots as the name suggests creates a regression line between two parameters and helps to visualize their linear relationships.
- There are various types of categorical plots:
  - a. lmplot
  - b. Regplot

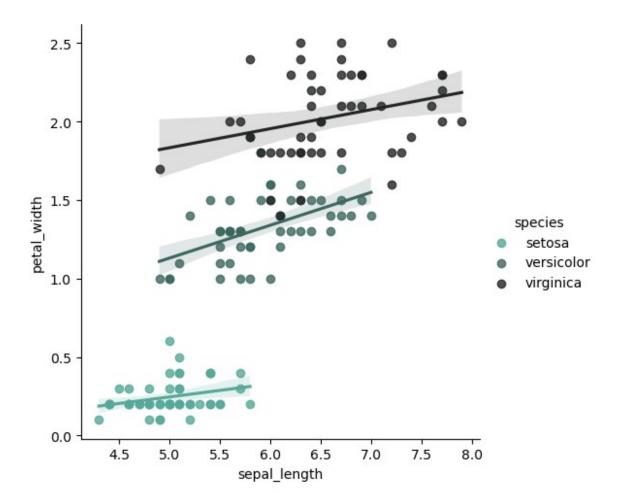
### 1. lmplot:

- Implot() method can be understood as a function that basically creates a linear model plot.
- It creates a scatter plot with a linear fit on top of it.
- Syntax:
  - seaborn.lmplot(x, y, data, hue=None, col=None, row=None,
    \*\*kwargs)

```
sns.lmplot(x='sepal_length', y='sepal_width', data=iris,
fit_reg=False)
plt.show()
```



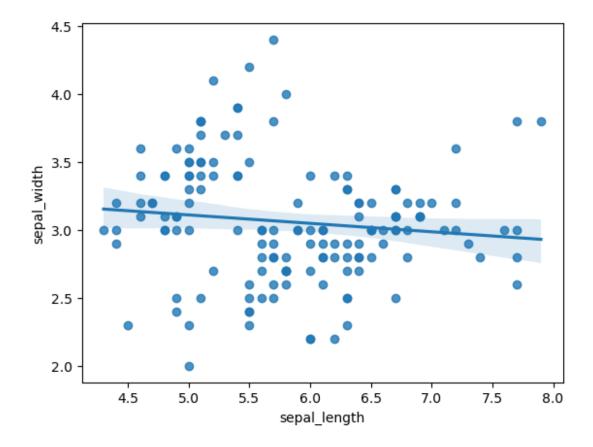
```
sns.lmplot(x='sepal_length', y='petal_width', data=iris,
hue='species', palette='dark:#5A9_r')
plt.show()
```



# 2. Regplot:

- regplot() method is also similar to Implot which creates linear regression model.
- Syntax:
  - seaborn.regplot( x, y, data=None, x\_estimator=None,
    \*\*kwargs)

```
sns.regplot(x='sepal_length', y='sepal_width', data=iris)
plt.show()
```



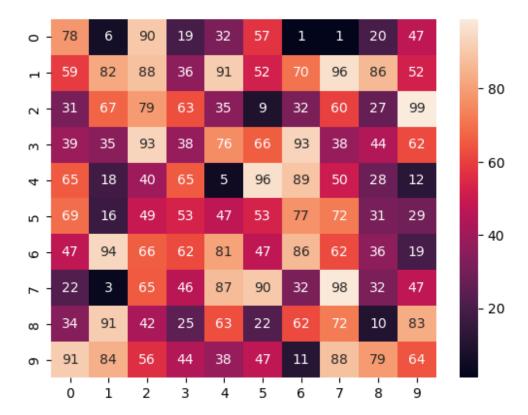
#### **Matrix Plots**

- A matrix plot means plotting matrix data, where color coded diagrams shows rows data, column data and values.
- It can shown using the heatmap and clustermap.
- There are various types of categorical plots:
  - a. Heatmap
  - b. Clustermap

## 1. Heatmap:

- Heatmap is defined as a graphical representation of data using colors to visualize the value of the matrix.
- In this, to represent more common values or higher activities brighter colors basically reddish colors are used and to represent less common or activity values, darker colors are preferred.
- It can be plotted using the heatmap() function.
- Syntax:
  - seaborn.heatmap(data, \*, vmin=None, vmax=None, cmap=None, linecolor='white', cbar=True, \*\*kwargs)

```
import numpy as np
import pandas as pd
data = np.random.randint(low=1, high=100, size=(10, 10))
hm = sns.heatmap(data=data, annot=True)
plt.show()
```



### 2. Clustermap:

- The clustermap() function of seaborn plots the hierarchically-clustered heatmap of the given matrix dataset.
- Clustering simply means grouping data based on relationship among the variables in the data.
- Syntax:
  - clustermap(data, \*, pivot\_kws=None, \*\*kwargs)

```
s1 = [100, 94, 56, 76, 81, 91, 51, 55, 72, 66, 60, 58]

s2 = [82, 81, 94, 96, 93, 84, 80, 82, 84, 86, 81, 78]

s3 = [65, 61, 66, 62, 67, 71, 69, 73, 68, 64, 66, 70]

s4 = [150, 140, 145, 151, 156, 152, 160, 165, 159, 149, 155, 162]

s5 = [75, 74, 76, 78, 80, 82, 85, 81, 77, 73, 75, 67]

s6 = [80, 75, 70, 72, 67, 65, 62, 63, 65, 60, 66, 69]

months= ["Jan", "Feb", "Mar", "Apr",
```

```
"May", "Jun", "Jul", "Aug", 
"Sep", "Oct", "Nov", "Dec"]
d1 = {"State1":s1,}
      "State2":s2,
      "State3":s3,
      "State4":s4,
      "State5":s5,
      "State6":s6};
df = pd.DataFrame(data=d1, index=months)
print(df)
print(df.columns)
print(df.index)
sns.clustermap(df)
plt.show()
     State1 State2
                       State3
                               State4
                                        State5
                                                 State6
        100
                  82
                                   150
Jan
                           65
                                             75
                                                      80
                                             74
                                                      75
Feb
         94
                  81
                           61
                                   140
          56
                  94
                           66
                                   145
                                             76
                                                      70
Mar
                  96
                           62
                                   151
                                             78
                                                      72
Apr
          76
                  93
                                   156
                                                      67
May
         81
                           67
                                             80
Jun
         91
                  84
                           71
                                   152
                                             82
                                                      65
                  80
Jul
         51
                           69
                                   160
                                             85
                                                      62
         55
                  82
                           73
                                   165
                                             81
                                                      63
Aug
         72
                  84
                           68
                                   159
                                             77
                                                      65
Sep
                  86
                           64
                                   149
                                             73
                                                      60
0ct
         66
                  81
                           66
                                   155
                                             75
                                                      66
Nov
         60
Dec
          58
                  78
                           70
                                   162
                                             67
                                                      69
Index(['State1', 'State2', 'State3', 'State4', 'State5', 'State6'],
dtype='object')
Index(['Jan', 'Feb', 'Mar', 'Apr', 'May', 'Jun', 'Jul', 'Aug', 'Sep',
'Oct',
        'Nov', 'Dec'],
      dtype='object')
```

