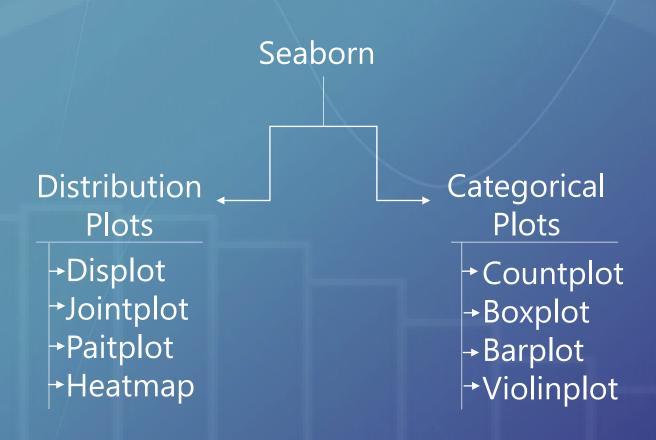
# DATA VISUALISATION USING SEABORN



## INTRODUCTION

Seaborn is a Python library for 2D and advanced visualization of certain dataframe. It helps to explore and analyse a dataset in more in-depth manner.

There are two varieties of plotting that can be done in seaborn- 1)Distribution Plotting and 2)Categorical Plotting.





### Importing Library

```
In [3]: import seaborn as sns
         import warnings
         warnings.filterwarnings('ignore')
In [4]: # Loading inbuilt dataframe from seaborn module
         df=sns.load dataset("penguins")
         df.head()
Out[4]:
            species
                       island
                              bill_length_mm bill_depth_mm
                                                           flipper_length_mm
                                                                             body_mass_g
                                                                                              sex
              Adelie
                    Torgersen
                                        39.1
                                                      18.7
                                                                       181.0
                                                                                    3750.0
                                                                                             Male
             Adelie
                                        39.5
                                                                       186.0
                                                                                    3800.0 Female
         1
                    Torgersen
                                                      17.4
              Adelie Torgersen
                                        40.3
                                                      18.0
                                                                       195.0
                                                                                    3250.0 Female
         2
              Adelie
                    Torgersen
                                        NaN
                                                      NaN
                                                                        NaN
                                                                                      NaN
                                                                                             NaN
              Adelie Torgersen
                                                      19.3
                                                                       193.0
                                                                                    3450.0 Female
                                        36.7
```

### Distribution Plots

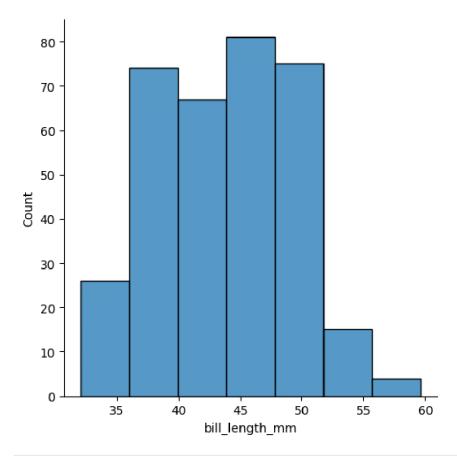
### **Displot**

Displot is used to analyse distribution of column features. It will give a histogram of column parameter. kde here means Kernel Density plot over histogram, it is set to True by default. Bins are the number of buckets you want to create for the column distribution. discrete is used to get rid of unnecessary gaps in the plot. You can also use the kind variable inside this to change the kind of plot. Also there are some other variables like hue, pallette, legend, rug, etc.

```
In [5]: #distplot(dataframe, kde=True/False, bins, stat, ...)
        sns.displot(df['bill_length_mm'], kde=False, bins=7)
```

Out[5]: <seaborn.axisgrid.FacetGrid at 0x1d0511e10d0>

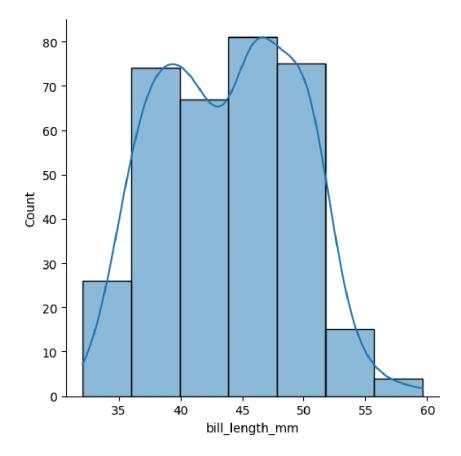




In [6]: sns.displot(df['bill\_length\_mm'], kde=True, bins=7)

Out[6]: <seaborn.axisgrid.FacetGrid at 0x1d0512082d0>





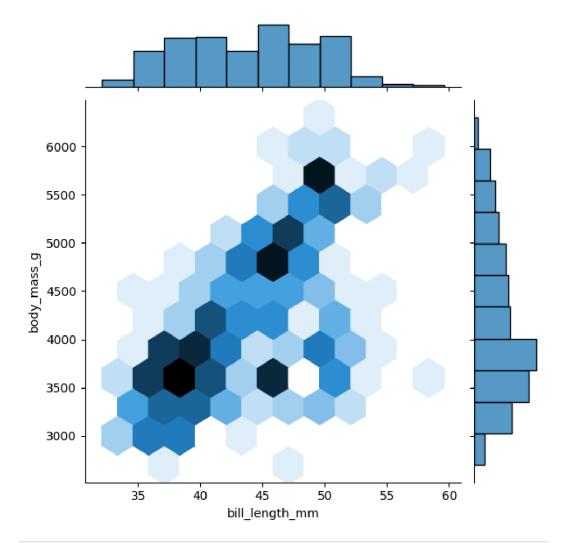
### Joinplot

A join plot allows to study the relationship between multiple features in comparison to each other. The central chart display the correlation of the columns. It is usually a scatterplot, a hexbin plot, a 2D histogram or a 2D density plot.

```
In [7]: #jointplot(data, x_axis, y_axis, hue, kind, palette,...)
        sns.jointplot(x='bill_length_mm', y='body_mass_g', data=df, kind='hex')
```

Out[7]: <seaborn.axisgrid.JointGrid at 0x1d05131fed0>

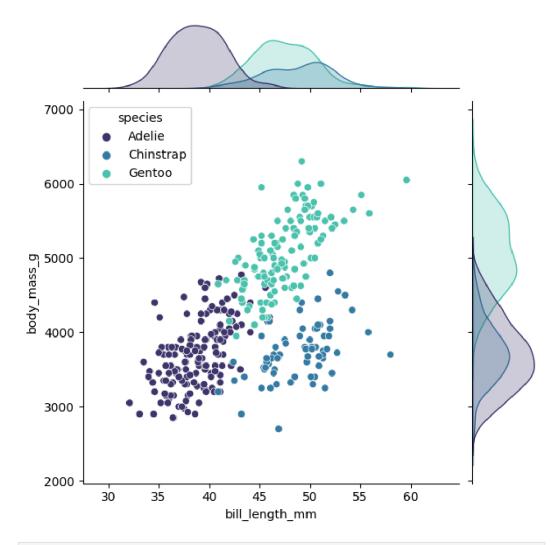




In [8]: sns.jointplot(x='bill\_length\_mm', y='body\_mass\_g', data=df, hue='species', p

Out[8]: <seaborn.axisgrid.JointGrid at 0x1d053a35890>

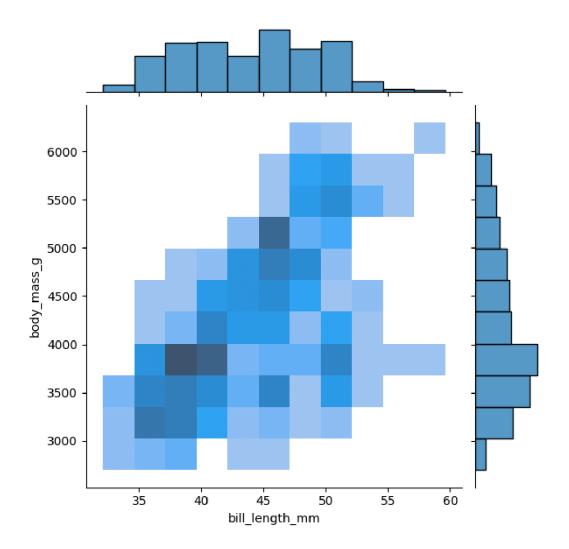




In [9]: sns.jointplot(x='bill\_length\_mm', y='body\_mass\_g', kind='hist', data=df)

Out[9]: <seaborn.axisgrid.JointGrid at 0x1d053c94cd0>





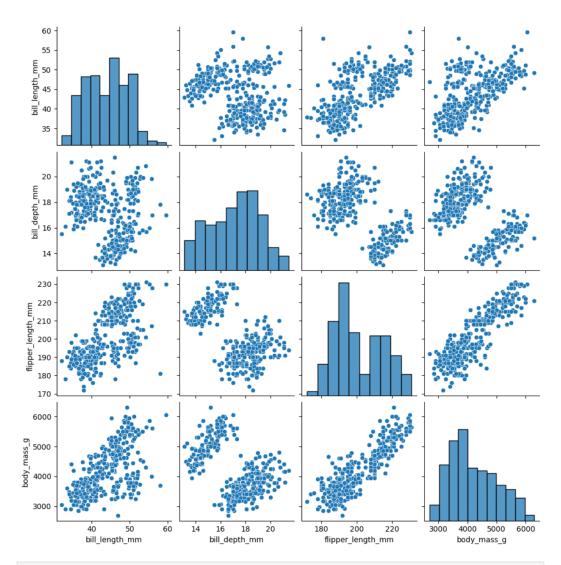
### Pairplot

Pairs plots shows the variations of all the parameters with respect to each other.

```
In [10]: #pairplot(data, hue, palette, vars, kind, ...)
sns.pairplot(df)
```

Out[10]: <seaborn.axisgrid.PairGrid at 0x1d053c92910>

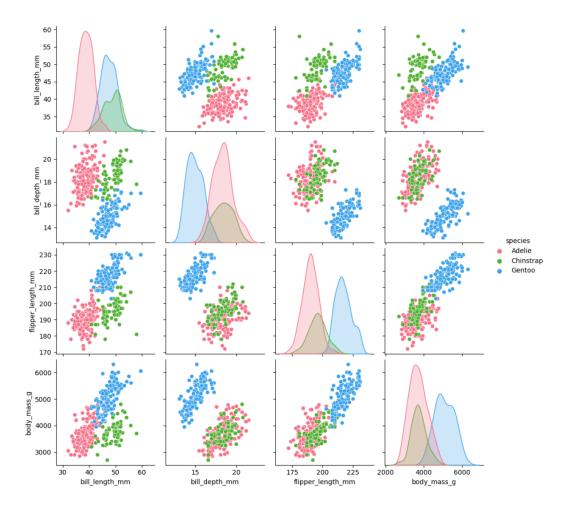




In [11]: sns.pairplot(df, palette='husl', hue='species')

Out[11]: <seaborn.axisgrid.PairGrid at 0x1d055846f50>





### Correlation / Heatmap

A correlation is a heatmap that uses colored cellscorrelation matrix (table) between two discrete dimensions or event types. It is very important in Feature Selection

In [12]: df.corr()

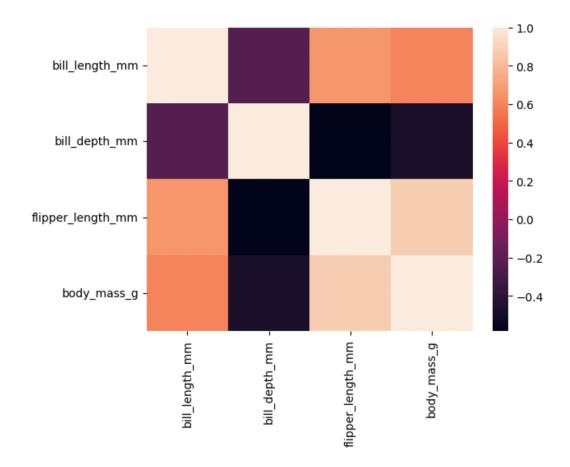
Out[12]:

	bili_lengtn_mm	biii_deptn_mm	tilpper_lengtn_mm	body_mass_g
bill_length_mm	1.000000	-0.235053	0.656181	0.595110
bill_depth_mm	-0.235053	1.000000	-0.583851	-0.471916
flipper_length_mm	0.656181	-0.583851	1.000000	0.871202
body_mass_g	0.595110	-0.471916	0.871202	1.000000

In [22]: sns.heatmap(df.corr())

Out[22]: <Axes: >



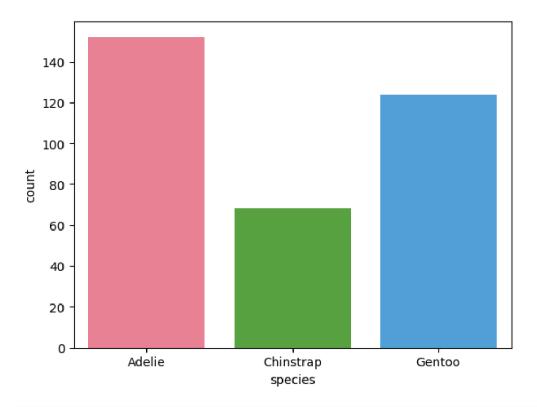


### **Categorical Plots**

### Countplot

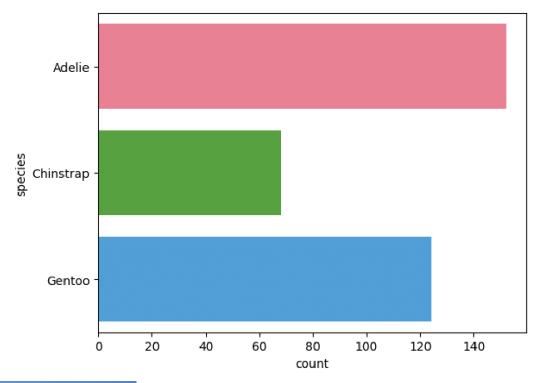
Count plot shows count of each variable in form of bars.





In [23]: sns.countplot(data=df, y='species', palette='husl')

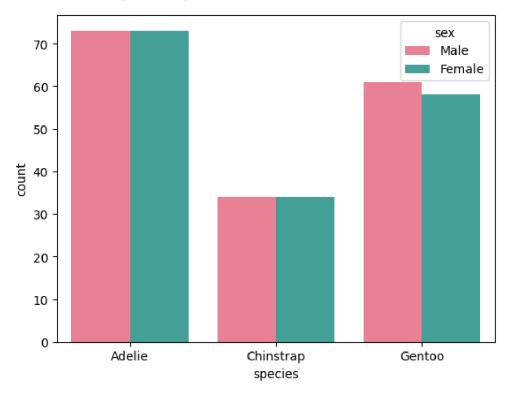
Out[23]: <Axes: xlabel='count', ylabel='species'>





```
In [24]: sns.countplot(data=df, x='species', hue='sex', palette='husl')
```

Out[24]: <Axes: xlabel='species', ylabel='count'>

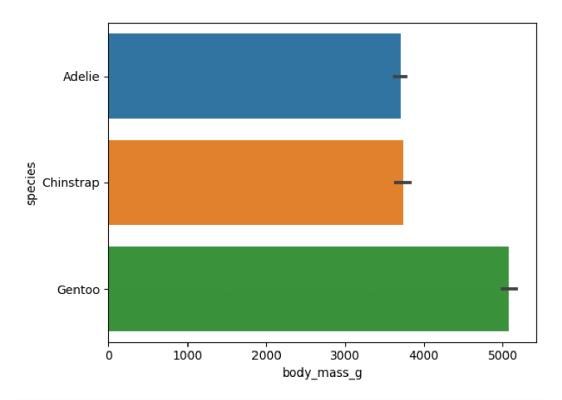


### **Barplot**

It represents an estimate of central tendency in form of bars

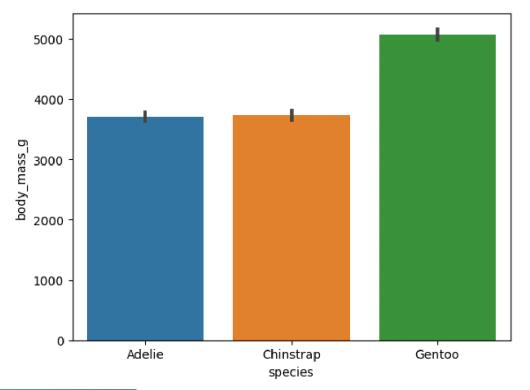
```
In [26]: #barplot(data, x_axis, y_axis, hue, order, ...)
sns.barplot(x='body_mass_g', y='species', data=df)
Out[26]: <Axes: xlabel='body_mass_g', ylabel='species'>
```





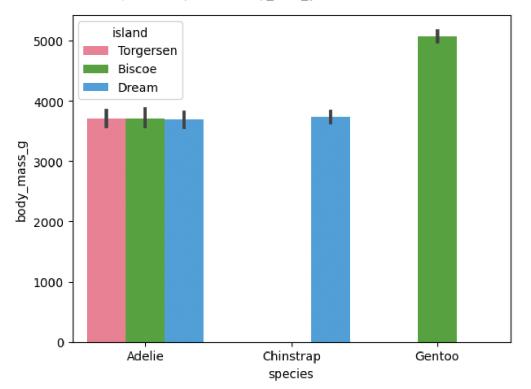
In [27]: sns.barplot(y='body\_mass\_g', x='species', data=df)

Out[27]: <Axes: xlabel='species', ylabel='body\_mass\_g'>





```
In [29]: sns.barplot(y='body_mass_g', x='species', data=df, hue='island', palette='hu
Out[29]: <Axes: xlabel='species', ylabel='body_mass_g'>
```



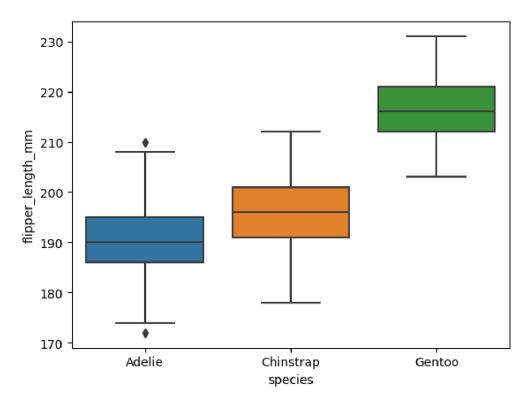
### Box plot

A box plot is a graph that presents information from a five-number summary i.e., lower\_value, 25%, median, 75%, upper\_value. It helps to find out the outliers in a dataframe.

```
In [30]: #boxplot(data, x_axis, y_axis, hue, order, orient, color, palette, ...)
sns.boxplot(x='species', y='flipper_length_mm', data=df)
```

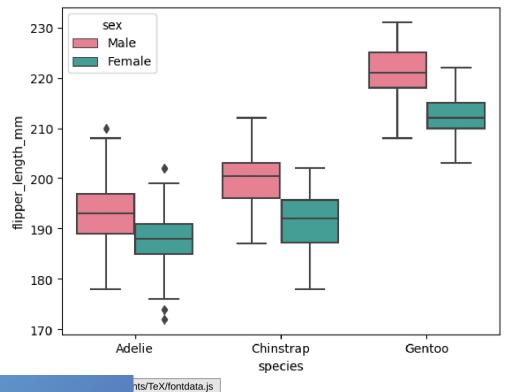
Out[30]: <Axes: xlabel='species', ylabel='flipper\_length\_mm'>





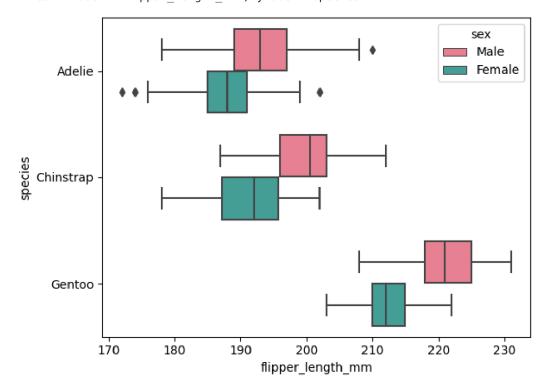
In [32]: sns.boxplot(x='species', y='flipper\_length\_mm', data=df, hue='sex', palette=

Out[32]: <Axes: xlabel='species', ylabel='flipper\_length\_mm'>





```
In [33]: sns.boxplot(y='species', x='flipper_length_mm', data=df, hue='sex', palette=
Out[33]: <Axes: xlabel='flipper_length_mm', ylabel='species'>
```

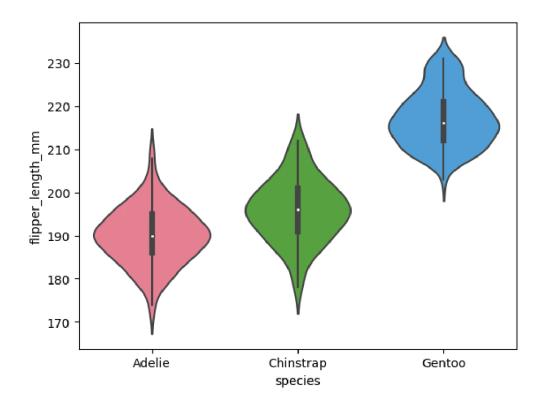


### Violin Plot

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

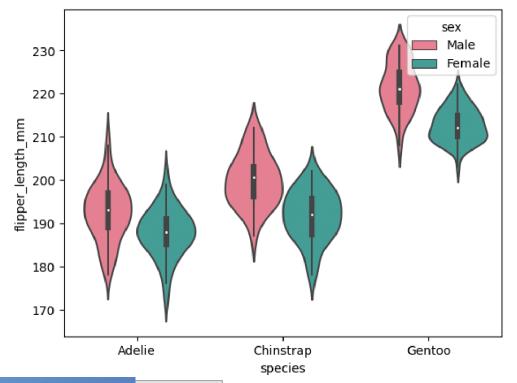
```
In [35]: sns.violinplot(x='species', y='flipper_length_mm', data=df, palette='husl')
Out[35]: <Axes: xlabel='species', ylabel='flipper_length_mm'>
```





In [36]: sns.violinplot(x='species', y='flipper\_length\_mm', data=df, hue='sex', palet

Out[36]: <Axes: xlabel='species', ylabel='flipper\_length\_mm'>







## What's your take from this???

