



# DATA VISUALISATION IN PYTHON

## CHEATSHEET

### Why Is Data Visualisation an Important Concept ?

- Because it help us understand distribution, trend, relationship, comparison and composition of data values
- It helps decision makers to quickly examine large piles of data and discover the hidden pattern/insights

**“ BEAUTY OF AN ART LIES IN THE MESSAGE IT CONVEYS ”**

### WHAT IS REQUIRED TO MAKE VISUALISATION IN PYTHON ?

#### MATPLOTLIB

Python based plotting library offers matplotlib with a complete 2D support along with limited 3D graphic support. It is useful in producing publication quality figures in interactive environment across platforms.

#### SEABORN

Being based on matplotlib, seaborn offers various features such as built in themes, color palettes, functions and tools to visualize univariate, bivariate, linear regression, matrices of data, statistical time series etc which lets us to build complex visualizations.

# Sample Data Set Used For The VISUALISATION Show Below

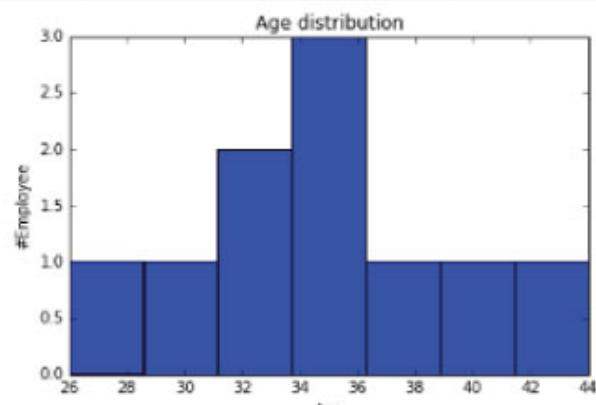
EMPID	Gender	Age	Sales	BMI	Income
E001	M	34	123	Normal	350
E002	F	40	114	Overweight	450
E003	F	37	135	Obesity	169
E004	M	30	139	Underweight	189
E005	F	44	117	Underweight	183
E006	M	36	121	Normal	80
E007	M	32	133	Obesity	166
E008	F	26	140	Normal	120
E009	M	32	133	Normal	75
E010	M	36	133	Underweight	40

## Import Data Set:

```
import matplotlib.pyplot as plt  
import pandas as pd  
df=pd.read_excel("E:/First.xlsx", "Sheet1")
```

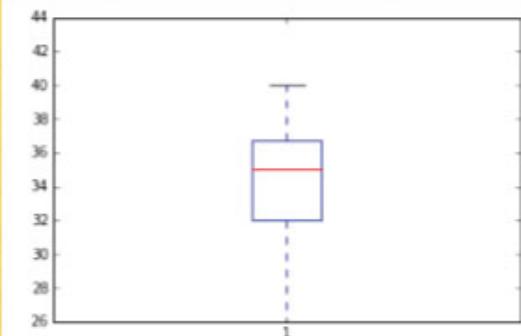
## Histogram

```
fig=plt.figure()  
ax = fig.add_subplot(1,1,1)  
ax.hist(df['Age'],bins = 7) # Here you can  
play with number of bins Labels and Tit  
plt.title('Age distribution')  
plt.xlabel('Age')  
plt.ylabel('#Employee')  
plt.show()
```



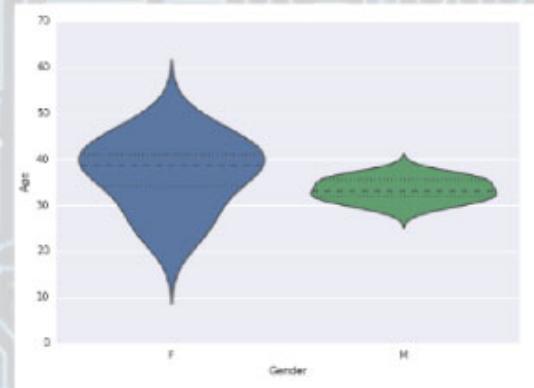
## Box Plot

```
import matplotlib.pyplot as plt  
import pandas as pd  
fig=plt.figure()  
ax = fig.add_subplot(1,1,1)  
x.boxplot(df['Age'])  
plt.show()
```



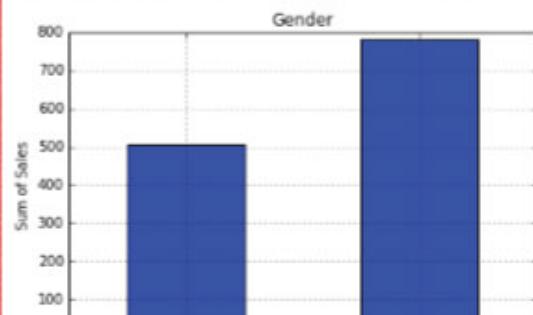
## Violin Plot

```
import seaborn as sns  
sns.violinplot(df['Age'], df['Gender'])  
#Variable Plot  
sns.despine()
```



## Bar Chart

```
var = df.groupby('Gender').Sales.sum()  
#grouped sum of sales at  
Gender level  
fig = plt.figure()  
ax1 = fig.add_subplot(1,1,1)  
ax1.set_xlabel('Gender')
```



```

23 ax1.set_ylabel('Sum of Sales')
23 ax1.set_title("Gender wise Sum of Sales")
23 var.plot(kind='bar')

```

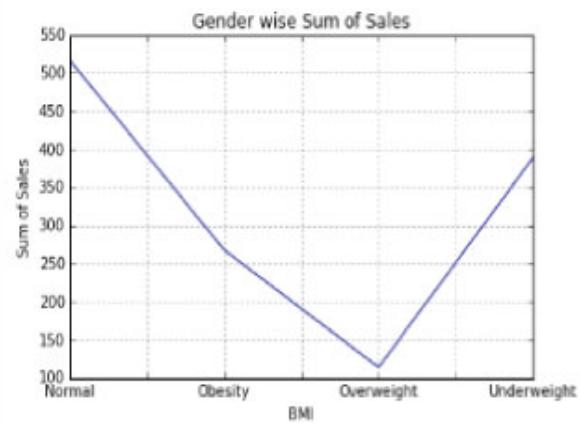


## Line Chart

```

var = df.groupby('BMI').Sales.sum()
fig = plt.figure()
ax1 = fig.add_subplot(1,1)
ax1.set_xlabel('BMI')
ax1.set_ylabel('Sum of Sales')
ax1.set_title("BMI wise Sum of Sales")
var.plot(kind='line')

```

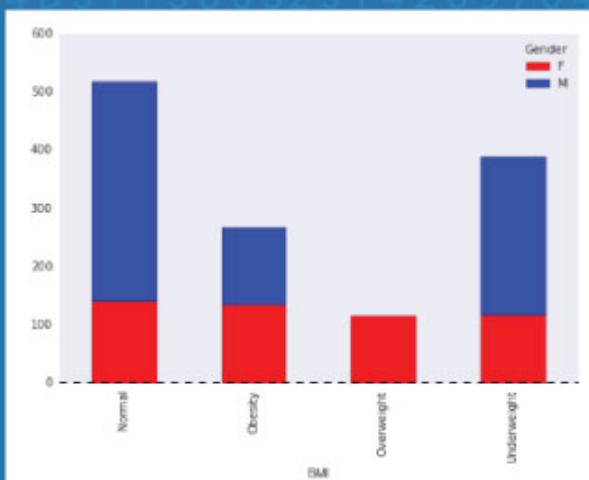


## Stacked Column Chart

```

var = df.groupby(['BMI','Gender']).Sales.sum()
var.unstack().plot(kind='bar',stacked=True, color=['red','blue'], grid=False)

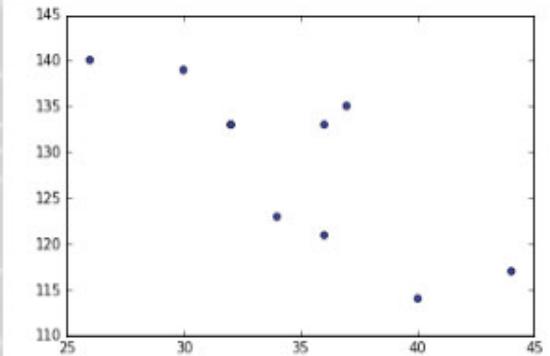
```



## Scatter Plot

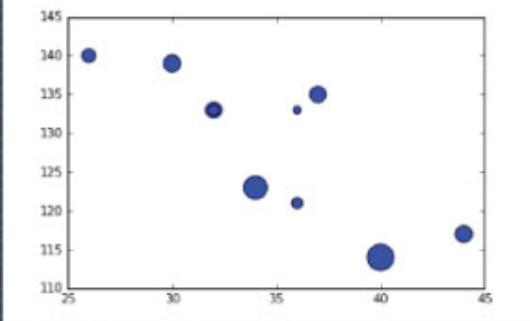
## Scatter Plot

```
fig = plt.figure()  
ax = fig.add_subplot(1,1,1)  
ax.scatter(df['Age'],df['Sales'])  
plt.show()
```



## Bubble Plot

```
fig = plt.figure()  
ax = fig.add_subplot(1,1,1)  
ax.scatter(df['Age'],df['Sales'], s=df['Income'])  
plt.show()
```



## Pie Chart

```
var=df.groupby(['Gender']).sum().stack()  
temp=var.unstack()  
type(temp)  
x_list = temp['Sales']
```

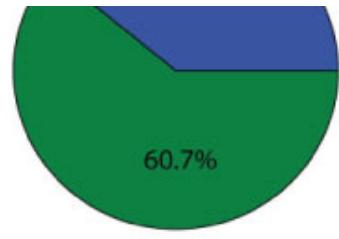
Pastafarianism expenses



```

label_list = temp.index
pyplot.axis("equal") #The pie chart
is oval by default. To make it a
circle use pyplot.axis("equal")
plt.pie(x_list,labels=label_list,autopct="%1.1f%%")
plt.title("Pastafarianism expenses")
plt.show()

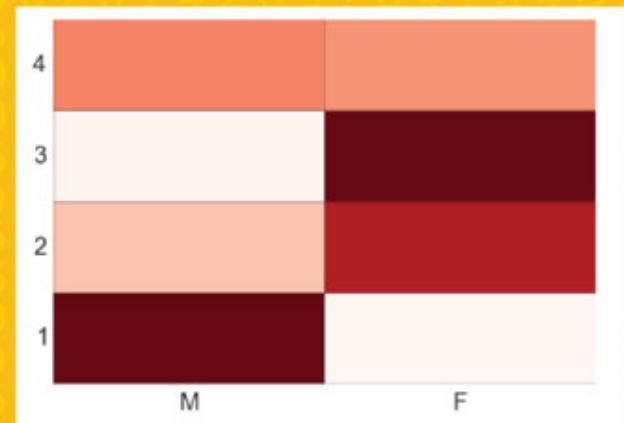
```



```

import numpy as np
data = np.random.rand(4,2)
rows = list('1234') #rows
categories columns =
list('MF') #column categories
fig,ax=plt.subplots()
ax.pcolor(data,cmap=plt.cm.Reds,edgecolors='k')
ax.set_xticks(np.arange(0,2)+0.5)
ax.set_yticks(np.arange(0,4)+0.5)
ax.xaxis.tick_bottom()
ax.yaxis.tick_left()
ax.set_xticklabels(categories,minor=False,fontsize=20)
ax.set_yticklabels(rows,minor=False,fontsize=20)
plt.show()

```



To view the complete guide on data visualisation in python  
**visit here : <http://bit.ly/1FjTkRF>**

