

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 SEABORN

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.

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. **EMPID** Sales BMI Gender Income E001 M 123 Normal 350 114 Overweight 450 E002 E003 135 Obesity 169 139 Underweight 189 183 117 Underweight

121 Normal

133 Obesity

140 Normal

133 Normal

133 Underweight

80

120

75

40

Set Used For The VISUALISATION **Show Below**

Sample Data

import matplotlib.pyplot as plt import pandas as pd

E007

E008

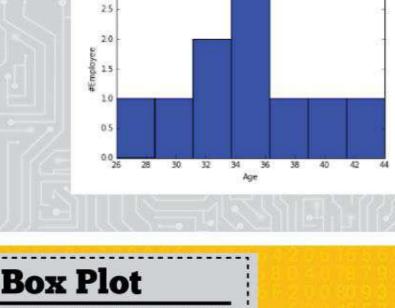
E009

E010

df=pd.read_excel("E:/First.xlsx", "Sheet1")

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()



Age distribution

ax = fig.add_subplot(1,1,1)

plt.show() import seaborn as sns sns.violinplot(df['Age'], df['Gender'])

importmatplotlib.pyplot as plt

import pandas as pd

x.boxplot(df['Age'])

fig=plt.figure()

#Variable Plot

sns.despine()

Gender level

fig = plt.figure()



var = df.groupby('Gender').Sales.sum() #grouped sum of sales at

ax1.set_title("Gender wise Sum of Sales")

ax1 = fig.add_subplot(1,1,1) ax1.set_xlabel('Gender') ax1.set_ylabel('Sum of Sales')

300

200

var = df.groupby('BMl').Sales.sum() fig = plt.figure() ax1 = fig.add_subplot(1,1,1)

ax1.set_xlabel('BMI')

of Sales")

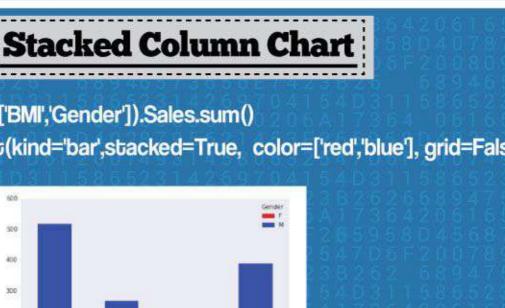
var.plot(kind='bar')

var.plot(kind='line')

ax1.set_ylabel('Sum of Sales')

ax1.set_title("BMI wise Sum

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



Bubble Plot

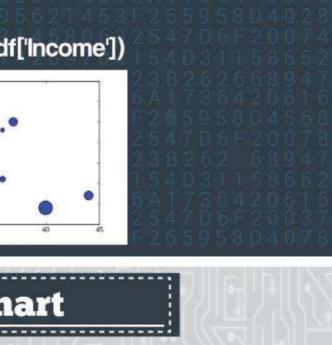
fig = plt.figure() ax = fig.add_subplot(1,1,1) ax.scatter(df['Age'],df['Sales'], s=df['Income']) plt.show()

fig = plt.figure()

plt.show()

ax = fig.add_subplot(1,1,1)

ax.scatter(df['Age'],df['Sales'])



Pastafarianism expenses

39.7%

x_list = temp['Sales']

pyplot.axis("equal") #The pie chart is oval by default. To make it a circle use pyplot.axis("equal")

var=df.groupby(['Gender']).sum().stack()

temp=var.unstack()

label_list = temp.index

type(temp)

plt.show()

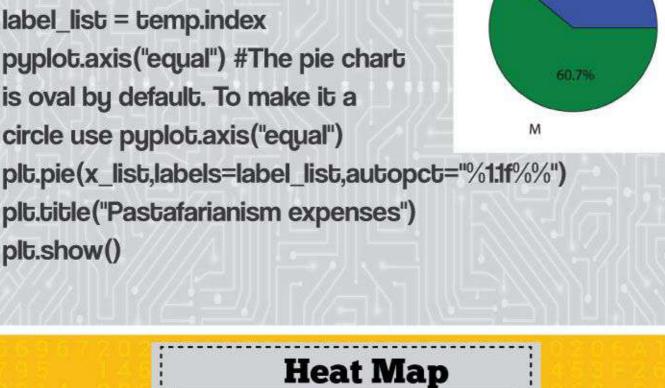
import numpy as np

data = np.random.rand(4,2)

rows = list('1234') #rows

categories columns =

fig,ax=plt.subplots()



list('MF') #column categories

3

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(columns,minor=False,fontsize=20) ax.set_yticklabels(rows,minor=False,fontsize=20) plt.show())



Analytics Vidhya Learn Everything About Analytics