DATA HANDLING AND VISUALIZATION LABSHEETS

NAME: RAHUL.S

ROLL NO: 20201ISB0026

LABSHEET-1 INTRODUCTION TO NUMPY

```
b=np.array([1,2,3])
add=np.add(a,b) add
 ⊕ array([2, 4,
6])
a=np.array([5,10,20])
b=np.array([4,8,10])
sub=np.subtract(a,b) sub
 10])
a=np.array([5,10,20])
b=np.array([4,8,10])
sub=np.multiply(a,b) sub

    array([ 20, 80, 80, 80])

200])
a=np.array([5,7,9]
b=np.array([4,5,6]
)
sub=np.mod(a,b) sub
 3])
a=np.array([1,2,3])
b=np.array([1,2,3])
add=np.power(a,b) add
\implies array([1, 4, 27])
Series creation
import
        pandas
                as pd
                          import
                                   numpy
                                           as
                                               np
data=np.array(['a','b','c','d'])
s=pd.Series(data) print(s)
0 ==
      а
1
      b
2
      C
3
      d
```

import numpy as np a=np.array([1,2,3])

dtype: object

Series with index

```
import pandas as pd import numpy as np
data=np.array(['a','b','c','d'])
s=pd.Series(data,index=[101,102,103,104]) print(s)

101    a
        102    b
        103    c
        104    d
        dtype:
object
```

Series with Dictionary

Series with Dictionary with index

```
import pandas as pd import numpy as np
data={'a': 0.,'b': 1.,'c': 2.}
s=pd.Series(data,index=['b','c','d','a']) print(s)

b
    1.0 c
    2.0 d
    NaN a
    0.0

dtype: float64
```

Create Series from Scalar

Retrieving data from the zeroth position

```
20201ISB0026
import pandas as pd
s= pd.Series([1,2,3,4,5],index=['a','b','c','d','e']) print(s[0])
  Ŧ
import pandas as pd
pd.Series([100,101,102,103,104,105,106,107,108,109,110],index=['a','b','c','d','e','f','
','h','i','j','k'])
print(s[:3])
101
     102
C
dtype: int64
import pandas as pd
pd.Series([100,101,102,103,104,105,106,107,108,109,110],index=['a','b','c','d','e','f','
g ','h','i','j','k']) print(s[2:8])
  ∓ c
     102 d
     103 e
     104 f
     105 g
     106 h
     107
dtype: int64
Using lable value
import pandas as pd
pd.Series([100,101,102,103,104,105,106,107,108,109,110],index=['a','b','c','d','e','f','
','h','i','j','k'])
print(s['a']) = 100
import pandas as pd
pd.Series([100,101,102,103,104,105,106,107,108,109,110],index=['a','b','c','d','e','f','
','h','i','j','k'])
print(s[['a','e','i','d']])
  ∄ a
     100 e
     104 i
     108 d
```

Data Frames

dtype: int64

```
import pandas as pd
df=pd.read_csv("/content/nyc_weather.csv")
```

Create data frame with empty data

Ricky 29

```
import pandas as pd df=pd.DataFrame()
 print(df)
Index: []
Create data frame from list
 import
             pandas
                         as
                                  pd
 data=[1,2,3,4,5]
 df=pd.DataFrame(data) print(df)
Ŧ
       0
0 1
1 2
2 3
3 4
4 5
 import pandas as pd
 data=[['Alex',10],['Bob',12],['Clarke',13]]
 df=pd.DataFrame(data,columns=['Name','Age']) print(df) Name Age
Alex
Bob
      10
      12
Clarke 13
 import pandas as pd
data=[['Dha',21, 10001,'A'],['Sha',23, 10002,'B'],['Dee',22, 10003,'C']]
df=pd.DataFrame(data,columns=['Name','Age','Rollno','Sec'],dtype=float) print(df)
Ŧ
        Name Age Rollno Sec
0 Dha 21.0 10001.0 A
1 Sha 23.0 10002.0 B
 2 Dee 22.0 10003.0 C
 <ipython-input-31-f22448152035>:3: FutureWarning: Could not cast to float64, falling
 back to object. This behavior is deprecated. I
 df=pd.DataFrame(data,columns=['Name','Age','Rollno','Sec'],dtype=float)
Cretae data frame from Dictionary
 import pandas as pd
 data={'Name':['Tom','Jack','Steve','Ricky'],'Age':[23,25,22,29]
 }
 df=pd.DataFrame(data,index=['rank1','rank2','rank3','rank4'])
 print(df)
   ₹ Name Age
            rank1
      Tom
            rank2
      Jack 25
            rank3
      Steve 22 rank4
```

LABSHEET-2 WORKING WITH PANDAS

```
import pandas as pd
  def load_data(): df_all =
pd.read_csv('/content/train.csv')
return df_all.loc[:300,['Survived','Pclass','Sex','Cabin','Embarked']].dropna()
df=load_data()
 df.head()
 Ŧ
        Survived Pclass Sex
                          Cabin
                                    Embarked
               1
                     male
                          C30
                                    S
 1
                     female D33
                                    С
               1
 9
         1
               3
                     male
                          E121
                                    S
 10
        1
               1
                     female B22
                                    S
 14
        0
               1
                     male
                            B51 B53 B55 S
FINDING DUPLICATE ROWS
df.Cabin.duplicated()
False
9
       False
       False
10
       False
14
. . .
       False
271
278
       False
286
       False
False
False
Name: Cabin, Length: 80, dtype: bool
df.duplicated()
False
False
9
       False
10
       False
       False
14
       False
271
278
       False
286
       False
False
Length: 80, dtype: bool
df.duplicated(subset=['Survived', 'Pclass', 'Sex'])
False
False
       False
9
       True
10
14
       True
271
       True
```

278

True

286 True299 True300 True

Length: 80, dtype: bool

COUNTING DUPLICATES AND NON DUPLICATES

df.duplicated(subset=['Survived','Pclass','Sex']).sum() = 70 (~df.duplicated()).sum() = 77 EXTRACTING DUPLICATE ROWS USING LOC

df.loc[df.duplicated(), :]

_ _+		Survived	Pclass	Sex	Cabin	Embarked
	138	1	2	female	F33	S
	169	1	1	female	B77	S
	237	1	1	female B	96 B98	S

USING KEEP

df.loc[df.duplicated(keep='first'), :]

∓÷		Survived	Pclass	Sex	Cabin	Embarked	
	138	1	2 f	emale	F33	S	
	169	1	1 female		B77	S	
	237	1	1	female B	96 B98	S	

df.loc[df.duplicated(keep='last'), :]

₹		Survived	Pclass	Sex	Cabin	Embarked
	36	1	1	female	B77	S
	77	1	1	female B	96 B98	S
	134	1	2	female	F33	S

df.loc[df.duplicated(keep=False), :]

-						
		Survived	Pclass	Sex	Cabin	Embarked
	36	1	1 f	emale	B77	S
	77	1	1 female B96		6 B98	S
	134	1	2 f	emale	F33	S
	138	1	2 female		F33	S
	169	1	1 f	emale	B77	S
	237	1	1 f	emale B9	6 B98	S

DROPPING DUPLICATED ROWS

df.drop_duplicates()

		Survived	Pclass	Sex	C	abin	Embark	ed
	0	0	1		male	C3	60	S
	1	1	1		female	D3	3	С
	9	1	3		male	E1	21	S
	10	1	1		female	В2	2	S
	14	0	1	male	B51 B53 B55			S
	271	1	1	male		C93		s
	278	0	1	male		C111		С
	286	1	1	male		C148		С
	299	1	1		female	D2		S
	300	1	2		male	F2	2	S

77 rows × 5 columns

df.drop_duplicates(keep=False)

	Survived	Pclass	Sex	Cabin	Embarked	
0	0	1	male	C30	S	
1	1	1	female	D33	С	
9	1	3	male	E121	S	
10	1	1	female	B22	S	
14	0	1	male	B51 B53 B55	S	
271	1	1	male	C93	S	
278	0	1	male	C111	С	
286	1	1	male	C148	С	
299	1	1	female	D21	S	
300	1	2	male	F2	S	

2 Pclass 891 non-null int64 3 Sex 891 non-null object 4 Age 714 non-null float64

5 SibSp 891 non-null int64

4 Age

74 rows × 5 columns

LABSHEET-3 DATA CLEANING

```
import pandas as pd import numpy as np
df=pd.read_csv('/content/2,1 dataset titanic.csv')
cols=['Name','Ticket','Cabin'] df=df.drop(cols,axis=1)
df.info()
₹ <class 'pandas.core.frame.DataFrame'> RangeIndex: 891 entries, 0 to 890
Data columns (total 9 columns):
      # Column
                 Non-Null Count Dtype
          0 PassengerId
                         891 non-null int64
          1 Survived 891 non-null int64
```

6 Parch 891 non-null int64 7 Fare 891 non-null float64 8 Embarked 889 non-null object dtypes: float64(2), int64(5), object(2) memory usage: 62.8+ KB

```
20201ISB0026
df=df.dropna() df.info()
₹<class 'pandas.core.frame.DataFrame'> Int64Index: 712 entries, 0 to 890
Data columns (total 9 columns):
      # Column
                   Non-Null Count Dtype
           0 PassengerId
                          712 non-null int64
           1 Survived 712 non-null int64
           2 Pclass 712 non-null int64
           3 Sex
                   712 non-null object
                   712 non-null float64
           4 Age
           5 SibSp
                   712 non-null int64
           6 Parch
                  712 non-null int64
           7 Fare
                   712 non-null float64
           8 Embarked 712 non-null object dtypes: float64(2), int64(5), object(2) memory usage:
                  dummies=[] cols=['Pclass','Sex','Embarked'] for col in cols:
     dummies.append(pd.get_dummies(df[col]))
titanic_dummies= pd.concat(dummies,axis=1)
df= pd.concat((df,titanic_dummies), axis=1)
df= df.drop(['Pclass','Sex','Embarked'],axis=1)
df['Age'] = df['Age'].interpolate() print(df)
  Ŧ
           PassengerId Survived Age SibSp Parch
                                              Fare 1 2 3 female \
                               0 7.2500 0 0 1
           0 1
                   0 22.0 1
                                                0
           1 2
                   1 38.0
                          1
                                  0 71.2833 1 0 0
                                               1
                   1 26.0 0 0 7.9250 0 0 1
1 35.0 1 0 53.1000 1 0 0
           2 3
                                                1
           3 4
                                                1
           ... .. ... 885 886 0 39.0 0 5 29.1250 0 0 1 1
                         27.0 0 0
19.0 0 0
26.0 0 0
32.0 0 0
     886
           887
                    0
                                         13.0000 0 1 0
                                                        0
     887
           888
                    1
                                          30.0000 1 0 0
                                                        1
     889
           890
                    1
                                         30.0000 1 0 0
                                                        0
     890
           891
                    0
                                          7.7500 0 0 1
                                                        0
     male C Q S 0 1 0
     0 1
     1 0
            100
     2 0
            001
     3 0
            991
     4 1
            001
     885 0010
         1001
     887
          0001
         1100
     889
     890 1010
[712 rows x 14 columns]
MIN MAX SCALAR STANDARDIZATION
from sklearn.preprocessing import MinMaxScaler data=[[-1,2],[-0.5,6],[0,10],[1,18]]
scaler=MinMaxScaler()
```

```
from sklearn.preprocessing import MinMaxScaler data=[[-1,2],[-0.5,6],[0,10],[1,18]]
scaler=MinMaxScaler()
print(scaler.fit(data)) print(' ')
MinMaxScaler()
print(scaler.data_max_) print(' ')
print('scaler.transform(data)')
Implied the max of the ma
```

```
scaler.transform(data)
from numpy import asarray
from sklearn.preprocessing import StandardScaler data=asarray([[100,0.001],
[8,0.05],
[50,0.005],
[88,0.07],
[4,0.1]]
print(data)
scaler= StandardScaler()
scaled = scaler.fit_transform(data) print(scaled)
₹ [[1.0e+02 1.0e-03]
[8.0e+00 5.0e-02]
[5.0e+01 5.0e-03]
[8.8e+01 7.0e-02]
[4.0e+00 1.0e-01]]
[[ 1.26398112 -1.16389967]
[-1.06174414 0.12639634]
[ 0. -1.05856939]
[ 0.96062565 0.65304778]
[-1.16286263 1.44302493]]
from sklearn.preprocessing import MinMaxScaler data=[[-1,2],[-0.5,6],[0,10],[1,18]]
scaler=MinMaxScaler()
print(scaler.fit(data)) MinMaxScaler()
print(scaler.data_max_)
print('scaler.transform(data)')
scaler.transform(data)
```

LABSHEET-4 Z-SCORE NORMALIZATION

```
import
        numpy
                 as
                      np
                           data=
[1,2,2,2,3,1,1,15,2,2,2,3,1,1,2]
                            std=
mean=
          np.mean(data)
np.std(data)
print("mean of the dataset ids", mean) print("std is",
std) threshold=3 outlier=[] for i in data: z=(i-mean)/std
if z>threshold: outlier.append(i)
print("outlier in dataset is", outlier)
3.3598941782277745
outlier in dataset is [15]
```

20201ISB0026 LABSHEET-5 OUTLIER DETECTION WITH IQR

low_lim = Q1 - 1.5 * IQR up_lim = Q3 + 1.5 * IQR

20201ISB0026 **LABSHEET-6 MATPLOTLIB**

import pandas as pd import numpy as np
import matplotlib.pyplot as plt

df=pd.read_csv("/content/Toyota.csv", index_col = 0, na_values = ['??','???'])

df.info()

(Column Non-Null Count Dtype
0	Price 1436 non-null int64
1	Age 1336 non-null float64
2	KM 1421 non-null float64
3	FuelType 1336 non-null object
4	HP 1436 non-null object
5	MetColor 1286 non-null float64
6	Automatic 1436 non-null int64
7	CC 1436 non-null int64
8	Doors 1436 non-null object
9	Weight 1436 non-null int64

dtypes: float64(3), int64(4), object(3) memory usage: 123.4+ KB

df.dropna(axis=0,inplace=True) df

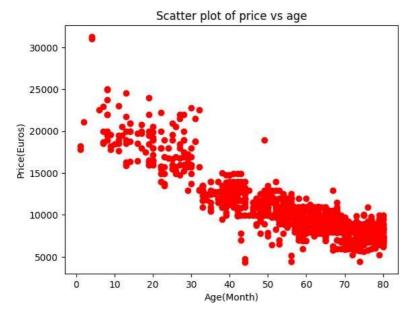
-													
<u>→</u> +		Price	Age	KM	FuelType	HP	MetC	olor	Auto	matic	CC	Doors	Weight
	0	13500 three		23.0 4698 1165	6.0	Dies	el	90		1.0	0		2000
	1	13750 3		23.0 7293 1165	37.0	Dies	el	90		1.0	0		2000
	3	14950 3		26.0 4800 1165	0.0	Dies	el	90		0.0	0		2000
	4	13750 3		30.0 3850 1170	0.0	Dies	el	90		0.0	0		2000
	5	12950 3		32.0 6100 1170	0.0	Dies	el	90		0.0	0		2000
	1423	7950 3		80.0 3582 1015	1.0	Petro	ol	86		0.0	1		1300
	1424	7750 3		73.0 3471 1015	7.0	Petro	ol	86		0.0	0		1300
	5	1429 8950 1065		78.0 2400	0.0	Petro	ol	86		1.0	1		1300

1430 8450 80.0 23000.0 Petrol 86 0.0 0 1300 3 1015 1435 6950 76.0 1.0 Petrol 110 0.0 0

1600 5 1114 1099 rows × 10 columns

SCATTER PLOT

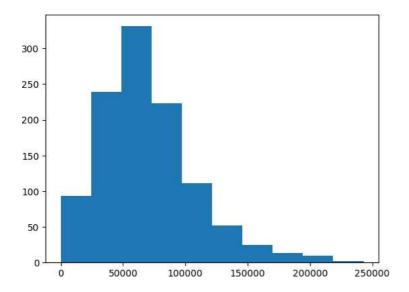
plt.scatter(df['Age'], df['Price'], c='red') plt.title('Scatter plot of price vs age')
plt.xlabel('Age(Month)') plt.ylabel('Price(Euros)') plt.show()



HISTOGRAM

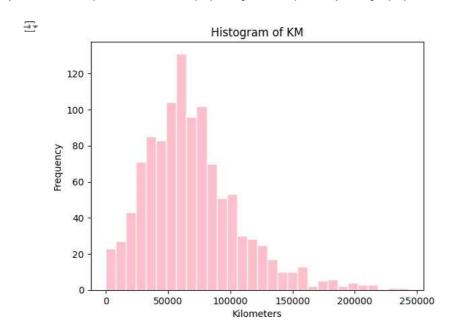
```
plt.hist(df['KM'])
```

```
1 (array([ 93., 239., 331., 223., 111., 52., 25., 13., 10., 2.]), array([1.000000e+00, 2.430090e+04, 4.860080e+04, 7.290070e+04, 9.720060e+04, 1.215005e+05, 1.458004e+05, 1.701003e+05, 1.944002e+05, 2.187001e+05, 2.430000e+05]),
Carray([ 93., 239., 331., 223., 111., 52., 25., 13., 10., 2.]), array([1.000000e+00, 2.430000e+04, 2.]), array([1.0000000e+00, 2.]), array([1.00000000e+00, 2.]), array([1.0000000e+00, 2.]), array([1.000000e+00, 2.]), ar
```



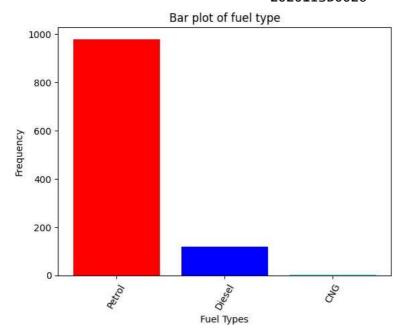
20201ISB0026 800 600 200 Diesel Petrol CNG

plt.hist(df['KM'],color='pink',edgecolor='white',bins=30) plt.title('Histogram of KM')
plt.xlabel('Kilometers') plt.ylabel('Frequency') plt.show()



BAR PLOT counts = [979,120,2] fueltype= ('Petrol','Diesel','CNG') index= np.arange(len(fueltype))

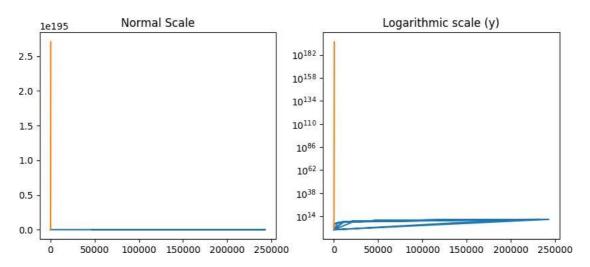
plt.bar(index,counts,color=['red','blue','cyan']) plt.title('Bar plot of fuel
type') plt.xlabel('Fuel Types') plt.ylabel('Frequency') plt.xticks(index,
fueltype, rotation= 60) plt.show()



LINE PLOT

```
fig, axes = plt.subplots(1, 2, figsize=(10,4)) x=df['KM']
axes[0].plot(x, x**2, x, np.exp(x)) axes[0].set_title("Normal Scale")
axes[1].plot(x, x**2, x, np.exp(x)) axes[1].set_yscale("log")
axes[1].set_title("Logarithmic scale (y)")
```

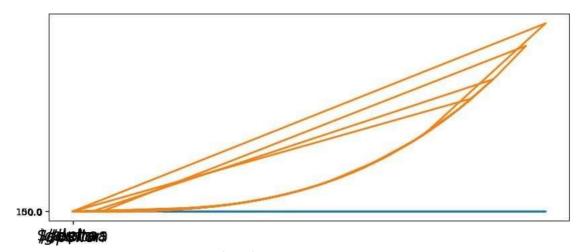
/usr/local/lib/python3.10/dist-packages/pandas/core/arraylike.py:396:
RuntimeWarning: overflow encountered in exp result = getattr(ufunc, method)(*inputs, **kwargs) Text(0.5, 1.0, 'Logarithmic scale (y)')



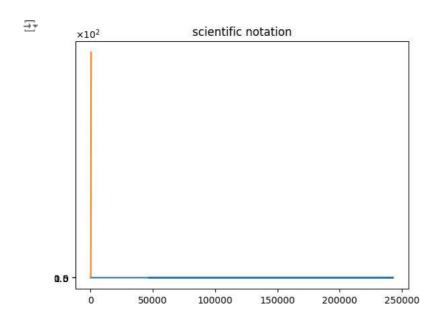
fig, ax = plt.subplots(figsize=(10,4)) x=df['KM']
ax.plot(x, x**2,x,x**3, lw=2) ax.set_xticks([1,2,3,4,5])
ax.set_xticklabels([r'\$/alphas',r'\$/beta\$',r'\$/gamma\$',r'\$/delta\$',
r'\$/epsilon\$'], fontsize=18) yticks=[0,50,100,150] ax.set_yticks(yticks)
ax.set_yticklabels(["\$%.1f\$" % y for y in yticks])

```
[Text(0, 0, '$0.0$'),
Text(0, 50, '$50.0$'),
Text(0, 100, '$100.0$'),
```

```
Text(0, 150, '$150.0$')]
```

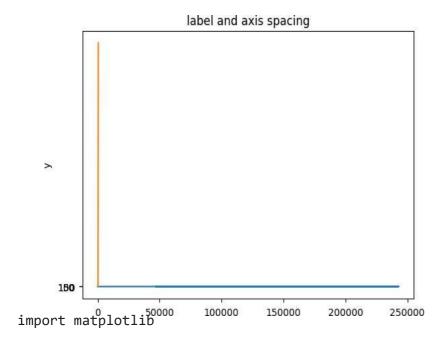


```
fig, ax= plt.subplots(1,1)
x=df['KM'] ax.plot(x, x**2, x,
np.exp(x))
ax.set_title("scientific notation") ax.set_yticks([0,50,100,150])
from matplotlib import ticker
formatter = ticker.ScalarFormatter(useMathText=True)
formatter.set_scientific(True) formatter.set_powerlimits((-1,1))
ax.yaxis.set_major_formatter(formatter)
```



```
import matplotlib
  matplotlib.rcParams['xtick.major.pad'] =
5 matplotlib.rcParams['ytick.major.pad']
= 5
  x =
df['KM']
fig, ax = plt.subplots(1, 1)
```

ax.plot(x, x**2, x, np.exp(x)) $ax.set_yticks([0, 50, 100,$ ax.set_title("label and axis spacing") ax.xaxis.labelpad = 5 ax.yaxis.labelpad ax.set_ylabel("x") ax.set_ylabel("y") plt.show() /usr/local/lib/python3.10/distpackages/pandas/core/arraylike.py:396: RuntimeWarning: overflow encountered in exp result = getattr(ufunc, method)(*inputs, **kwargs)



matplotlib.rcParams['xtick.major.pad'] = 3
matplotlib.rcParams['ytick.major.pad'] = 3

LABSHEET-7 INTERACTING WITH WEB API

```
import requests
pip install --upgrade 'library' 
→ Collecting library
Downloading Library-0.0.0.tar.gz (1.4 kB) Preparing metadata (setup.py) ... done
Building wheels for collected packages: library
Building wheel for library (setup.py) ... done
Created wheel for library: filename=Library-0.0.0-py3-none-any.whl size=2054
sha256=33e04a1cd46e5d3b86146af77a7e80978fe44edaeba4a Stored in directory:
/root/.cache/pip/wheels/e0/71/7d/b0e29b944e43374597cd4e3b88c85197001c9bfcd5dce191f4
Successfully built library
Installing collected packages: library Successfully installed library-0.0.0
r = requests.get('https://www.romexchange.com/')
r
r.status_code = 406 url =
'https://www.romexchange.com/'
headers = {'Content-type': 'application/json'}
ur
1

    'https://www.romexchange.com/'

header
url = 'https://www.romexchange.com/'
headers = {'User-Agent': 'XY', 'Content-type': 'application/json'} r = requests.get(url,
headers=headers)
url

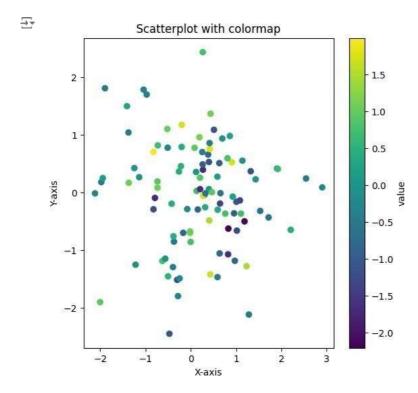
    'https://www.romexchange.com/'

header
r
```

LABSHEET-8 COLORMAPS

import pandas as pd import numpy as np import matplotlib.pyplot as plt

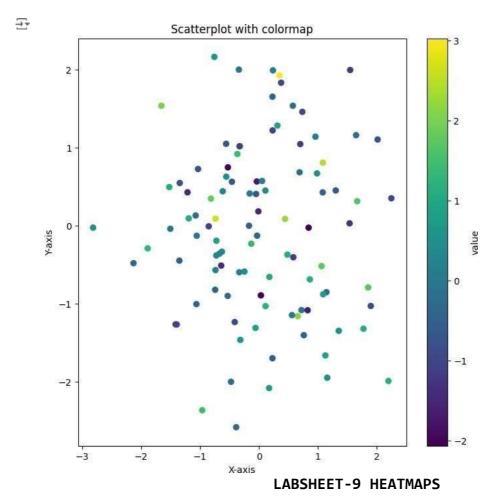
#sample datafame with multiple columns
data=pd.DataFrame({"x":np.random.randn(100),"y":np.random.randn(100),"value":np.random.
ran dn(100)}) #define the colormap and alpha values cmap="viridis" alpha=1 #create the
scatterplot plt.figure(figsize=(6,6))
plt.scatter(data["x"],data["y"],c=data["value"],cmap=cmap,alpha=alpha) #customize the
plot(optional)
plt.xlabel("X-axis") plt.ylabel("Y-axis")
plt.title("Scatterplot with colormap") plt.colorbar(label="value")
#show the plot plt.show()



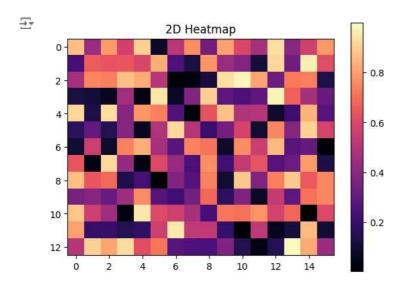
import pandas as pd import numpy as np
print(np.random.randn(100))

```
-1.72214971e+00 4.27042698e-01 -1.20346437e+00 2.83589309e-01 1.21334367e+00 4.14428011e-02 -1.48913563e+00 4.39560682e-01 -8.90366916e-01 -9.11298844e-01 3.62446399e-01 5.87632377e-01 1.22152619e+00 7.44396580e-01 1.75575979e+00 3.12178887e-01 -3.40512410e-01 -1.01818680e+00 4.62977518e-02 2.30443390e-01 -3.96879315e-01 1.20713778e+00 -1.20064064e+00 -9.12708432e-01 9.06172668e-01 7.05249075e-02 -9.42170303e-01 -8.52966288e-01 1.96198904e+00 3.61012540e-02 9.66762176e-01 -4.97875528e-01 2.78681896e-01 -1.16708383e+00 7.39087305e-01 1.27038245e+00 7.81304235e-01 -4.62440127e-01 1.00117969e+00 -9.07298230e-02 -1.95950298e-01 1.59291286e+00 -1.22572212e+00 -4.62563405e-01 5.41920487e-01 7.41261996e-01 1.42219990e+00 -9.65150475e-01] import pandas as pd import numpy as np import matplotlib.pyplot as plt
```

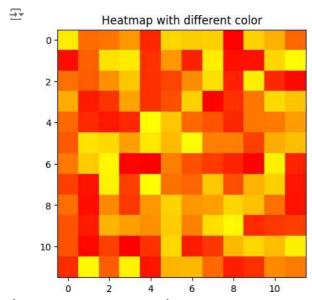
#sample datafame with multiple columns
data=pd.DataFrame({"x":np.random.randn(100),"y":np.random.randn(100),"value":np.random.
ran dn(100)}) #define the colormap and alpha values cmap="viridis" alpha=1 #create the
scatterplot plt.figure(figsize=(8,8))
plt.scatter(data["x"],data["y"],c=data["value"],cmap=cmap,alpha=alpha) #customize the
plot(optional)
plt.xlabel("X-axis") plt.ylabel("Y-axis")
plt.title("Scatterplot with colormap") plt.colorbar(label="value")
#show the plot plt.show()



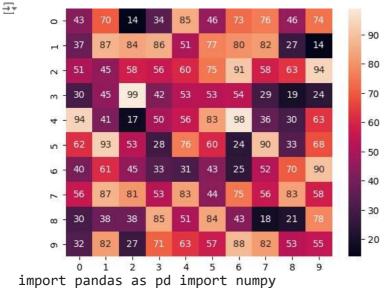
import matplotlib.pyplot as plt data= np.random.random((13,16))
plt.imshow(data,cmap="magma") plt.title("2D Heatmap") plt.colorbar()
plt.show()



import numpy as np
import matplotlib.pyplot as plt data=np.random.random((12,12)) plt.imshow(data,
cmap='autumn')
plt.title("Heatmap with different color") plt.show()



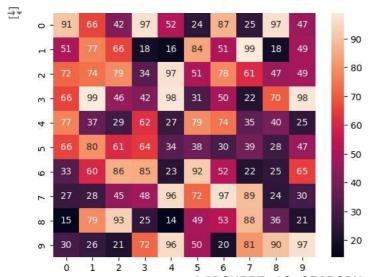
import numpy as np import
seaborn as sns import
matplotlib.pyplot as plt
data= np.random.randint(low=14,high=100, size=(10,10))
hm=sns.heatmap(data=data, annot=True) plt.show()



import pandas as pd import numpy
as np

df=pd.read_csv('/content/train.csv')

df=
np.random.randint(low=55,
high=60, size=(8,8))
hm=sns.heatmap(data=data, annot=True) plt.show()



LABSHEET-10 SEABORN COLOR PALLETTES

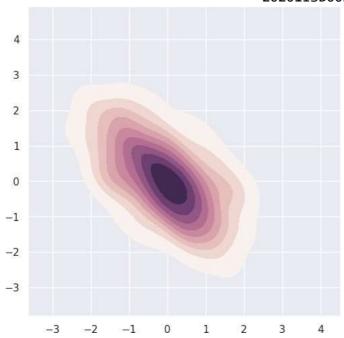
import numpy as np import pandas as pd
import matplotlib.pyplot as plt import seaborn as sns
%matplotlib inline

sns.set(rc={"figure.figsize": (6,6)})

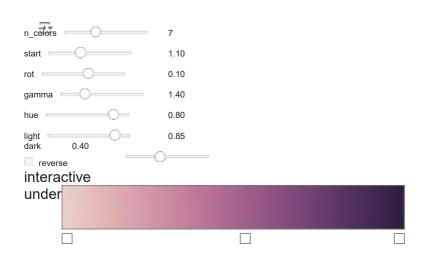
BUILDING COLOR PALLETTES

current_palette = sns.color_palette() sns.palplot(current_palette)

```
20201ISB0026
  Ŧ
sns.palplot(sns.color_palette("hls", 8))
  Ŧ
sns.palplot(sns.color_palette("husl", 8))
  Ŧ
sample_colors = ["windows blue", "amber", "greyish", "faded green", "dusty purple",
"pale red", "medium green", "denim blue"] sns.palplot(sns.xkcd_palette(sample_colors))
  Ŧ
sns.palplot(sns.color_palette("cubehelix", 8))
  Ŧ
sns.palplot(sns.cubehelix_palette(8))
  Ŧ
x,y = np.random.multivariate_normal([0,0], [[1,-.5],[-.5,1]], size=300).T sample_cmap =
sns.cubehelix palette(light=1, as cmap=True)
sns.kdeplot(x=x,y=y,cmap=sample_cmap, shade=True)
<ipython-input-16-534ef71d14c3>:3: FutureWarning:
`shade` is now deprecated in favor of `fill`; setting `fill=True`.
This will become an error in seaborn v0.14.0; please update your code.
sns.kdeplot(x=x,y=y,cmap=sample_cmap, shade=True)
<Axes: >
```



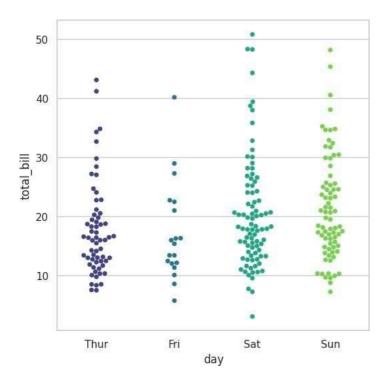
sns.choose_cubehelix_palette(as_cmap=True)



sns.set_style('whitegrid')

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l sns.swarmplot(x="day", y="total_bill", data=tips, palette="viridis")

<Axes: xlabel='day', ylabel='total_bill'>



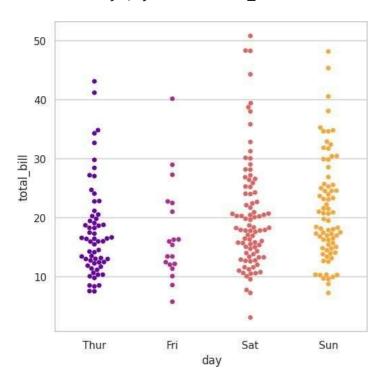
sns.set_style('whitegrid')

sns.swarmplot(x="day", y="total_bill", data=tips, palette="plasma") = <ipython-input-248931cda8de8a>:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l sns.swarmplot(x="day", y="total_bill", data=tips, palette="plasma")

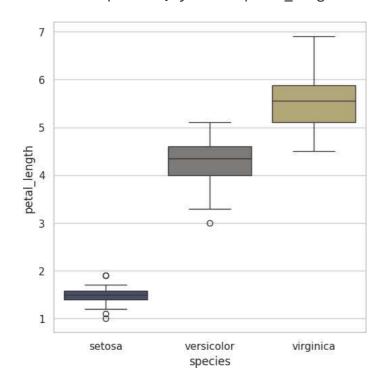


<Axes: xlabel='day', ylabel='total_bill'>



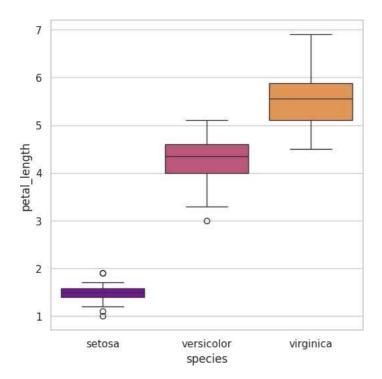
<Axes: xlabel='species', ylabel='petal_length'>

data=iris, palette="cividis")



Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l sns.boxplot(x="species", y="petal_length", data=iris, palette="plasma")

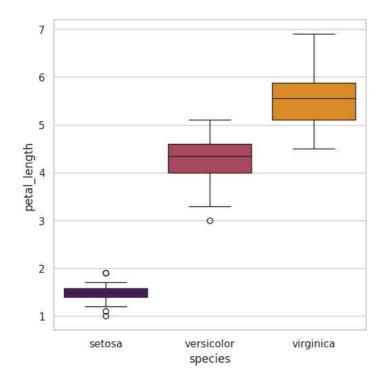
<Axes: xlabel='species', ylabel='petal_length'>



iris = sns.load_dataset("iris")
sns.boxplot(x="species", y="petal_length", data=iris, palette="inferno")

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l sns.boxplot(x="species", y="petal_length", data=iris, palette="inferno")

<Axes: xlabel='species', ylabel='petal_length'>



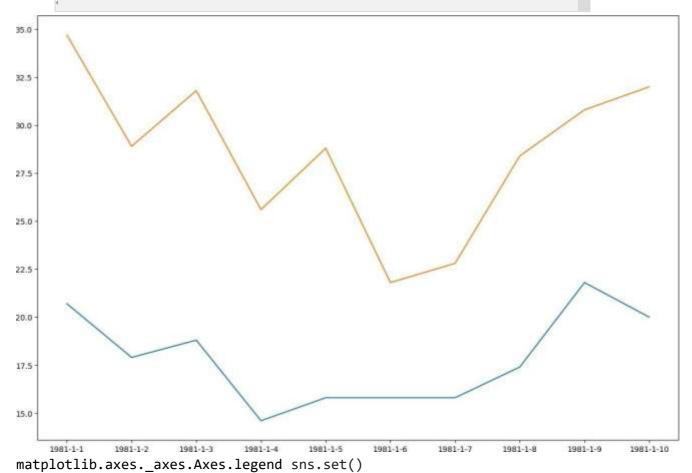
LABSHEET-11 MULTIVARIATE VISUALIZATION

Relational plots: relation b/w two variables categorical plots: categorical values are displayed distribution plots: examining univariate and bivariate distributions matrix plots: array of scatterplots Regression plots: emphasixe patterns in dataset during exploratory data analysis import numpy as np import pandas as pd import matplotlib.pyplot as plt from matplotlib.pyplot import figure import seaborn as sns %matplotlib inline dates = ['1981-1-1', '1981-1-2','1981-1-3','1981-1-4','1981-1-5','1981-1-6','1981-1-7','1981-1-8','1981-1-9','1981-1-10'] min temperature = [20.7,17.9,18.8,14.6,15.8,15.8,15.8,17.4,21.8,20.0] max_temperature = [34.7, 28.9, 31.8, 25.6, 28.8, 21.8, 22.8, 28.4, 30.8, 32.0]plt.subplots(nrows=1, fig,axes ncols=1, figsize=(15,10)) axes.plot(dates,min temperature, label='Min temperature')

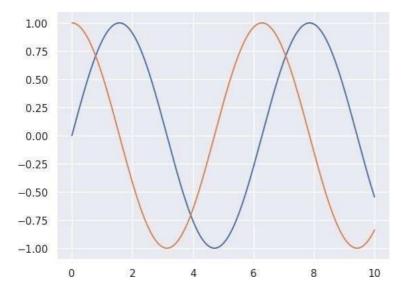
axes.plot(dates,max_temperature, label='Max temperature') axes.legend

```
def legend(*args, **kwargs)
/usr/local/lib/python3.10/dist-packages/matplotlib/axes/_axes.py
Place a legend on the Axes.
```

Call signatures::



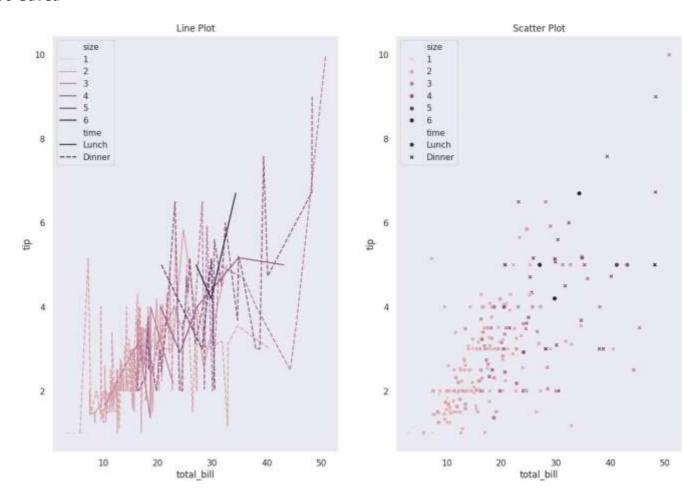
x = np.linspace(0,10,1000)
plt.plot(x, np.sin(x), x, np.cos(x))
== [<matplotlib.lines.Line2D at 0x7e3acaaaffa0>,
<matplotlib.lines.Line2D at 0x7e3acaae0040>]



sns.set(style="dark")

```
fig, ax = plt.subplots(ncols=2, nrows=1, figsize=(15,10)) df= sns.load_dataset("tips")
print(df.head())
sns.lineplot(x="total_bill", y="tip", hue="size", style= "time",
data=df,ax=ax[0]).set_title("Line Plot")
sct_plt = sns.scatterplot(x="total_bill", y="tip",
                                                                hue="size", style="time",
                                ax=ax[1]).set_title("Scatter
                                                                                       Plot")
data=df,
sct_plt.figure.savefig('Scatter_plot1.png') print('Plot Saved')
Ŧ
          total_bill tip sex smoker day time size
16.99 1.01 Female
                   No Sun Dinner
10.34 1.66
             Male
                   No Sun Dinner
21.01 3.50
             Male
                   No Sun Dinner
                                 3
                   No Sun Dinner
                                 2
23.68 3.31
            Male
24.59 3.61 Female
                   No Sun Dinner
```

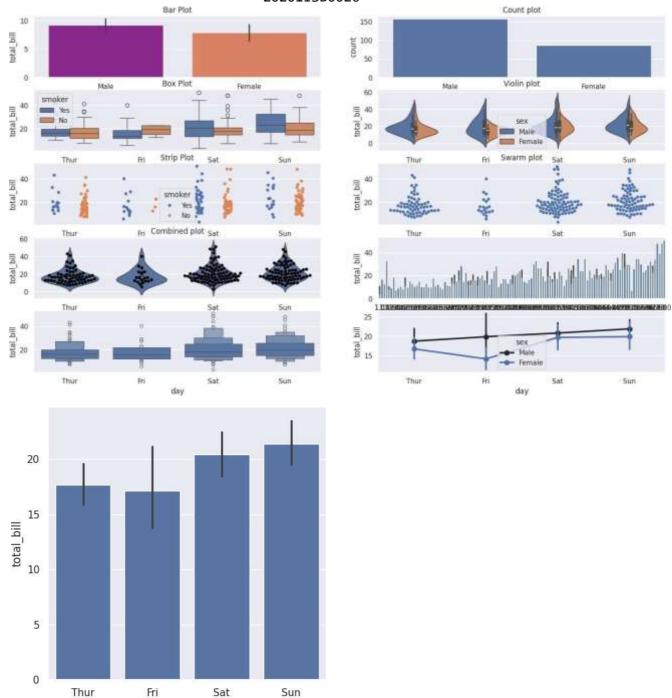
Plot Saved



```
sns.set_style('darkgrid')
fig, ax = plt.subplots(nrows=5, ncols=2) fig.set_size_inches(18.5, 10.5)
```

```
df=sns.load_dataset('tips')
```

```
data=df,
sns.barplot(x='sex', y='total_bill',
                                                 palette='plasma', estimator=
                                     Plot') sns.countplot(x='sex',
         ax=ax[0,0]).set_title('Bar
                                                                       data=df,
ax=ax[0,1]).set_title('Count plot')
sns.boxplot(x='day', y='total_bill', data=df, hue='smoker',
ax=ax[1,0]).set_title('Box Plot')
sns.violinplot(x='day', y='total_bill', data=df, hue='sex', split= True,
ax=ax[1,1]).set_title('Violin plot')
sns.stripplot(x='day', y='total_bill', data=df, jitter= True, hue='smoker',
dodge=True, ax=ax[2,0]).set_title('Strip Plot') sns.swarmplot(x='day', y='total_bill',
data=df, ax=ax[2,1]).set_title('Swarm plot')
sns.violinplot(x='day', y='total_bill', data=df, ax=ax[3,0])
sns.swarmplot(x='day',y='total_bill',data=df, color='black',
ax=ax[3,0]).set_title('Combined plot') sns.barplot(x='tip',y='total_bill', data=df,
ax=ax[3,1]
sns.boxenplot(x="day", y="total_bill", color="b", scale="linear", data=df, ax=ax[4,0])
sns.pointplot(x="day", y="total_bill", color="b", hue="sex", data=df, ax=ax[4,1])
79e72dcff921>:7: FutureWarning:
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0.
Assign the `x` variable to `hue` and set `l sns.barplot(x='sex', y='total_bill', data=df,
palette='plasma', estimator= np.std, ax=ax[0,0]).set title('Bar Plot') <ipython-input-6-</pre>
79e72dcff921>:24: FutureWarning:
The `scale` parameter has been renamed to `width_method` and will be removed in
v0.15. Pass `width_method='linear' for the same eff sns.boxenplot(x="day",
y="total_bill", color="b", scale="linear", data=df, ax=ax[4,0]) <ipython-input-6-
79e72dcff921>:26: FutureWarning:
Setting a gradient palette using color= is deprecated and will be removed in v0.14.0. Set
`palette='dark:b'` for the same effect. sns.pointplot(x="day", y="total bill", color="b",
hue="sex", data=df, ax=ax[4,1])
<seaborn.axisgrid.FacetGrid at 0x7e3ac3b802e0>
```



sns.set_style('whitegrid')
#loading the dataset directly without any files df=sns.load_dataset('iris')
print(df.head())

```
sepal_length sepal_width petal_length petal_width species

5.1 3.5 1.4 0.2 setosa

4.9 3.0 1.4 0.2 setosa

4.7 3.2 1.3 0.2 setosa

4.6 3.1 1.5 0.2 setosa

5.0 3.6 1.4 0.2 setosa
```

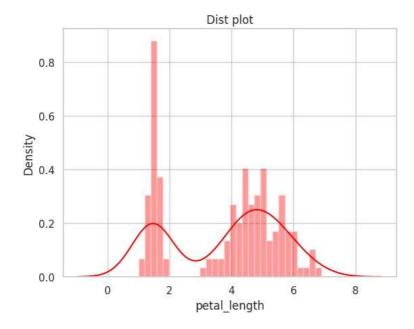
day

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sns.distplot(df['petal_length'], kde=True, color='red', bins=30).set_title('Dist plot')
Text(0.5, 1.0, 'Dist plot')



jointgrid = sns.JointGrid(x='petal_length', y='petal_width', data=df)
jointgrid.plot_joint(sns.scatterplot)
jointgrid.plot_marginals(sns.distplot) = /usr/local/lib/python3.10/distpackages/seaborn/axisgrid.py:1886: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
func(self.x, **orient_kw_x,

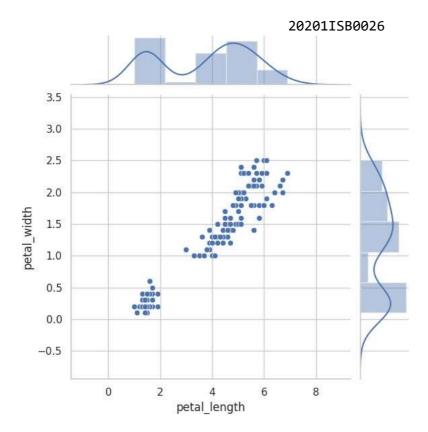
**kwargs)

/usr/local/lib/python3.10/dist-packages/seaborn/axisgrid.py:1892: UserWarning:

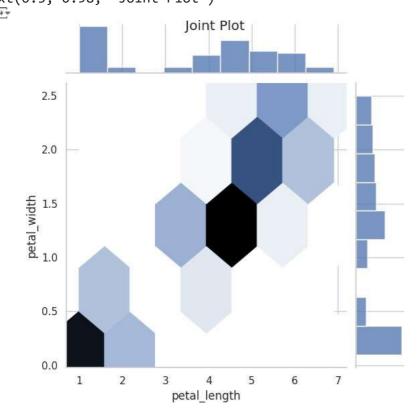
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms). For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

func(self.y, **orient_kw_y, **kwargs)
<seaborn.axisgrid.JointGrid at 0x7e3b00f8d120>



g=sns.jointplot(x='petal_length', y= 'petal_width', data=df, kind='hex')
g.fig.suptitle('Joint Plot')
Text(0.5, 0.98, 'Joint Plot')

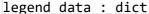


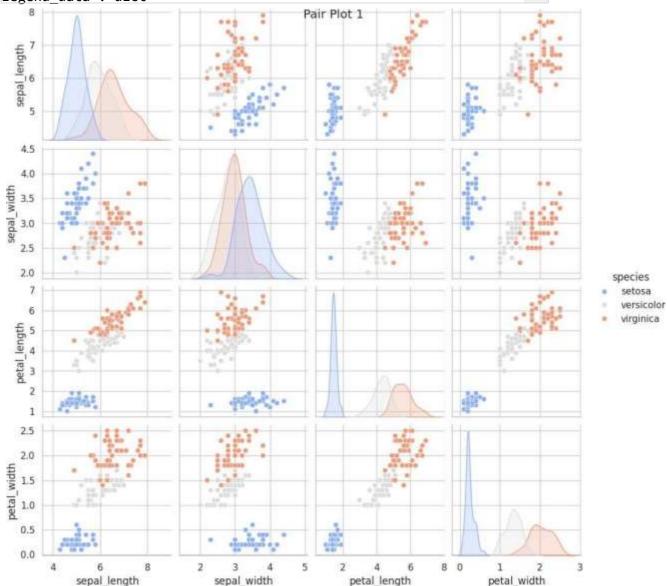
g=sns.pairplot(df, hue="species", palette= 'coolwarm') g.fig.suptitle("Pair Plot 1")
g.add_legend

```
seaborn.axisgrid.Grid.add_legend
def add_legend(legend_data=None, title=None, label_order=None,
adjust_subtitles=False, **kwargs)
/usr/local/lib/python3.10/dist-packages/seaborn/axisgrid.py
```

Draw a legend, maybe placing it outside axes and resizing the figure.

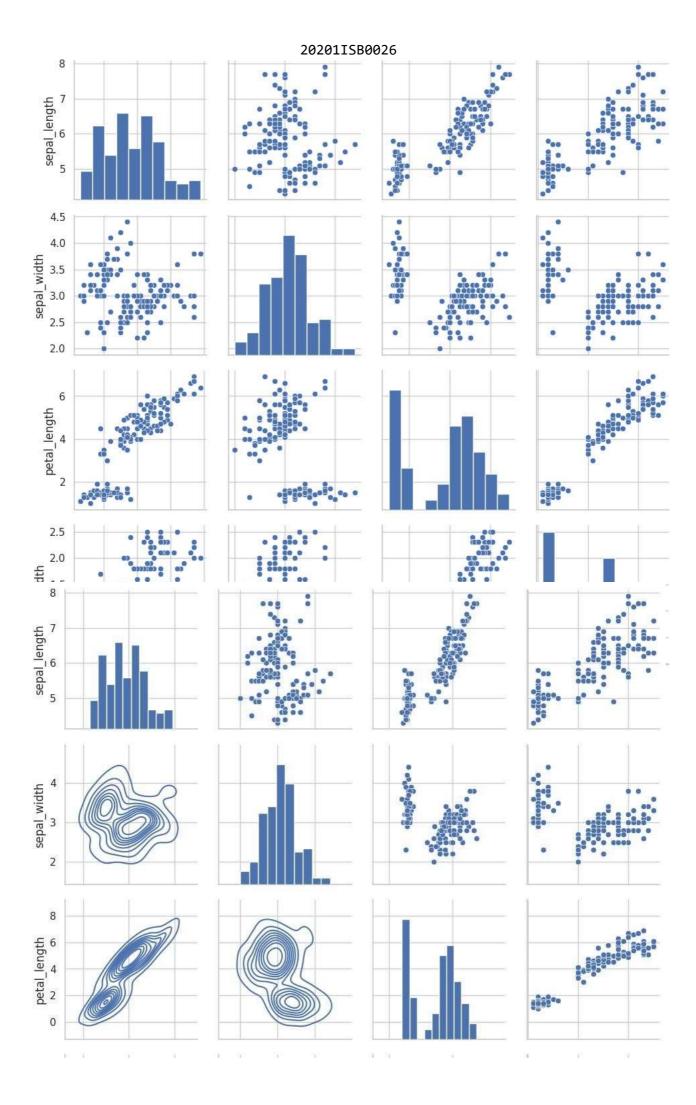
Parameters





```
pairgrid= sns.PairGrid(data=df)
pairgrid= pairgrid.map_offdiag(sns.scatterplot) pairgrid= pairgrid.map_diag(plt.hist)

pairgrid = sns.PairGrid(data=df)
pairgrid = pairgrid.map_upper(sns.scatterplot) pairgrid = pairgrid.map_diag(plt.hist)
pairgrid = pairgrid.map_lower(sns.kdeplot)
```



LABSHEET-12 TEXT VISUALIZATION

```
import pandas as pd
import matplotlib.pyplot as plt from wordcloud import WordCloud from wordcloud import
STOPWORDS
df= pd.read csv('/content/netflix titles.csv', usecols=['cast']) df.head()
cast
NaN
Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...
Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...
NaN
Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...
ndf=df.dropna() ndf.head()
cast
Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...
Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...
Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...
Kate Siegel, Zach Gilford, Hamish Linklater, H...
Vanessa Hudgens, Kimiko Glenn, James Marsden, ...
text= " ".join(item for item in ndf['cast']) print(text)
🛨 Ama Qamata, Khosi Ngema, Gail Mabalane, Thabang Molaba, Dillon Windvogel, Natasha
Thahane, Arno Greeff, Xolile Tshabalala, Getmore
stopwords = set(STOPWORDS)
wordcloud
                                 WordCloud(background_color="White").generate(text)
plt.imshow(wordcloud, interpolation= 'bilinear') plt.axis("off")
plt.margins(x=0, y=0) plt.show()
                       Richard
     Daniel
```

wordcloud = WordCloud(background_color="White", max_words=100, max_font_size=300, width=
800, height=500, colormap="magma").generate(te
plt.figure(figsize=(20,20)) plt.imshow(wordcloud, interpolation=
'bilinear') plt.axis("off") plt.margins(x=0, y=0) plt.show()



20201ISB0026 LABSHEET-13 TIME SERIES DATA

A time series is the series of data points listed in time order. A time series is a sequence of successive equal interval points in time. A time-series analysis consists of methods for analyzing time series data in order to extract meaningful insights and other useful characteristics of data. For performing time series analysis download stock data.csv import pandas as pd import numpy as np import matplotlib.pyplot as plt # reading the dataset using read csv df = pd.read_csv("/content/stock_data.csv", parse_dates=True, index_col="Date") # displaying the first five rows of dataset df.head() Open High Low Close Volume Name Date 2006-01-03 39.69 41.22 38.79 40.91 24232729 AABA 2006-01-04 41.22 41.90 40.77 40.97 20553479 AABA 2006-01-05 40.93 41.73 40.85 41.53 12829610 AABA 2006-01-06 42.88 43.57 42.80 43.21 29422828 AABA 2006-01-09 43.10 43.66 42.82 43.42 16268338 AABA

We have used the 'parse_dates' parameter in the read_csv function to convert the 'Date' column to the DatetimeIndex format. By default, Dates are stored in string format which is not the right format for time series data analysis.

Now, removing the unwanted columns from dataframe i.e. 'Unnamed: 0'.

Generate code with df | View recommended plots

deleting column
df=df.drop(columns='Name') print(df)

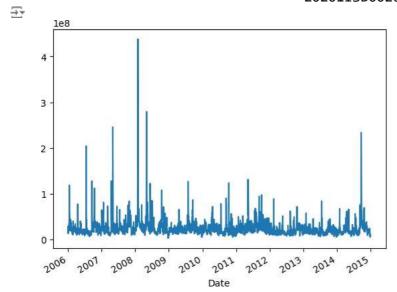
```
Date
2006-01-03 39.69 41.22 38.79 40.91 24232729
2006-01-04 41.22 41.90 40.77 40.97 20553479
2006-01-05 40.93 41.73 40.85 41.53 12829610
2006-01-06 42.88 43.57 42.80 43.21 29422828
2006-01-09 43.10 43.66 42.82 43.42 16268338
...
2014-12-23 51.46 51.46 49.93 50.02 15514036
2014-12-24 50.19 50.92 50.19 50.65 5962870
2014-12-26 50.65 51.06 50.61 50.86 5170048
2014-12-29 50.67 51.01 50.51 50.53 6624489 2014-12-30 50.35 51.27 50.35 51.22 10703455
```

[2263 rows $x \ 5 \ columns$]

Next steps:

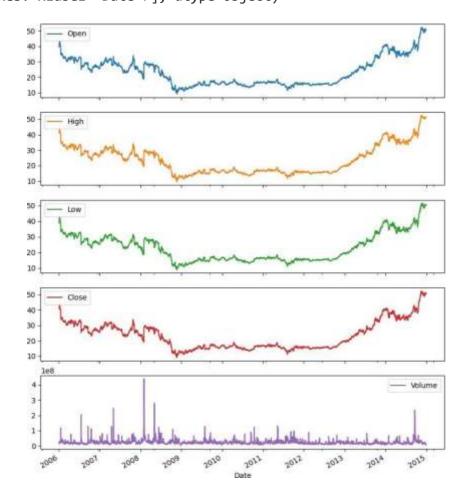
Example 1: Plotting a simple line plot for time series data.

df['Volume'].plot()
<Axes: xlabel='Date'>



Example 2: Now let's plot all other columns using subplot.

<Axes: xlabel='Date'>], dtype=object)



Resampling: Resampling is a methodology of economically using a data sample to improve the accuracy and quantify the uncertainty of a population parameter. Resampling for months or weeks and making bar plots is another very simple and widely used method of

finding seasonality. Here we are going to make a bar plot of month data for 2016 and 2017. Example 3:

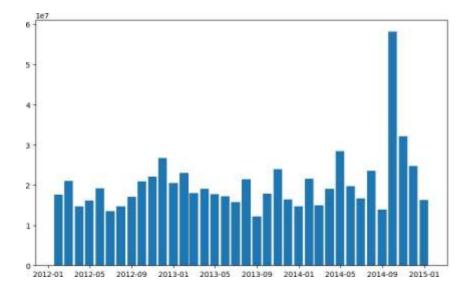
```
# Resampling the time series data based on monthly 'M' frequency df_month
= df.resample("M").mean() print(df_month)

# using subplot
fig, ax = plt.subplots(figsize=(10, 6))

# plotting bar graph
ax.bar(df_month['2012':'2014'].index, df_month.loc['2012':'2014', "Volume"],width=25,
align='center')
```

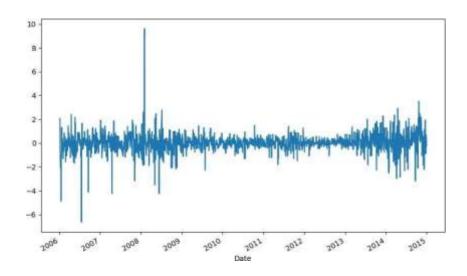
```
→÷
                     0pen
                                High
                                                     Close
                                                                  Volume
                                                                            Change
   Date
    2006-01-31 38.245500 38.694000 37.641500 38.113000 3.400594e+07 0.991442
   2006-02-28 33.141579 33.436842 32.627368 32.975789 2.329848e+07 0.996423
   2006-03-31 31.333478 31.696957 30.929130 31.218696 2.095522e+07 1.000390
   2006-04-30 32.383684 32.790000 31.914737 32.283158 2.200768e+07 1.001098
   2006-05-31 31.744545 32.175455 31.171364 31.517273 2.218047e+07 0.998535
   2014-08-31 36.836190 37.150000 36.545238 36.876667 1.396539e+07 1.003530
   2014-09-30 40.662857 41.270000 39.983810 40.671905 5.811769e+07 1.003005
   2014-10-31 41.253043 41.886087 40.784783 41.393913 3.210848e+07 1.005501
   2014-11-30 49.879474 50.553158 49.440000 50.151579 2.474402e+07 1.006233
    2014-12-31 50.359524 50.975714 49.852857 50.331905 1.623090e+07 0.999653
```

[108 rows x 6 columns] <BarContainer object of 36 artists>

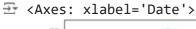


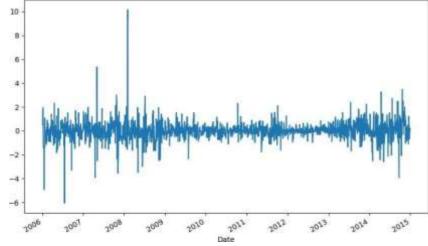
Differencing: Differencing is used to make the difference in values of a specified interval. By default, it's one, we can specify different values for plots. It is the most popular method to remove trends in the data.

df.Low.diff(2).plot(figsize=(10, 6))



df.High.diff(2).plot(figsize=(10, 6))





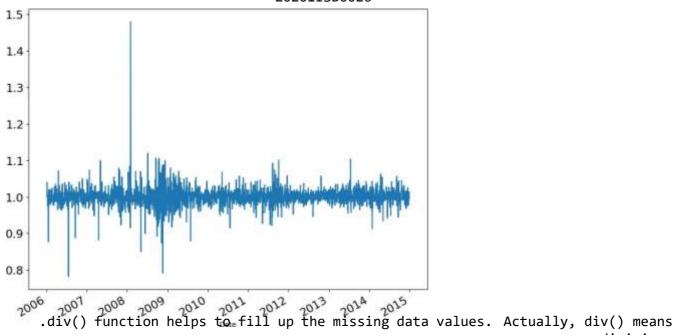
Plotting the Changes in Data

We can also plot the changes that occurred in data over time. There are a few ways to plot changes in data.

Shift: The shift function can be used to shift the data before or after the specified time interval. We can specify the time, and it will shift the data by one day by default.

That means we will get the previous day's data. It is helpful to see previous day data and today's data simultaneously side by side.

```
df['Change'] = df.Close.div(df.Close.shift())
df['Change'].plot(figsize=(10, 8), fontsize=16)
```



division.

If we take df. div(6) it will divide each element in df by 6.

We do this to avoid the null or missing values that are created by the 'shift()' operation.

df['Change'].plot(figsize=(10, 6))
<Axes: xlabel='Date'>

