DATA HANDLING AND VISUALIZATION LABSHEETS

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LABSHEET-1 INTRODUCTION TO NUMPY

```
add=np.add(a,b) add
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
200])
a=np.array([5,7,9]
b=np.array([4,5,6]
)
sub=np.mod(a,b) sub
31)
a=np.array([1,2,3])
b=np.array([1,2,3])
add=np.power(a,b) add
Series creation
import pandas as pd import
                              numpy
                                     as
                                         np
data=np.array(['a','b','c','d'])
s=pd.Series(data) print(s)
0 <del>∑</del> a
1
     b
2
     c
3
     d
```

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

b=np.array([1,2,3])

```
dtype: object
```

Series with index

Series with Dictionary

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

arrange a 0.0
b    1.0
c    2.0
dtype: float64
```

Series with Dictionary with index

Create Series from Scalar

Retrieving data from the zeroth position

```
20201ISB0027
  import pandas as pd
  s= pd.Series([1,2,3,4,5],index=['a','b','c','d','e']) print(s[0])
  1
    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])
    <u>∓</u> c
       102 d
       103 e
       104 f
       105 g
  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'])
  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']])
    100 e
       104 i
       108 d
       103
  dtype: int64
  Data Frames
  import pandas as pd
  df=pd.read_csv("/content/nyc_weather.csv")
```

Ricky 29

```
Create data frame with empty data
 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 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 10
Bob 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 23
           rank2
      Jack 25
            rank3
      Steve 22 rank4
```

271

278

True

True

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
 0
               1
                     male
                          C30
                                    S
               1
                     female D33
                                    С
 9
                                    S
         1
               3
                     male
                          E121
 10
        1
               1
                     female B22
                                    S
                            B51 B53 B55 S
 14
        0
               1
                     male
FINDING DUPLICATE ROWS
df.Cabin.duplicated()
False
False
9
       False
10
       False
14
       False
271
       False
278
       False
286
       False
False
False
Name: Cabin, Length: 80, dtype: bool
df.duplicated()
False
False
9
       False
       False
10
14
       False
. . .
       False
271
       False
278
       False
286
False
Length: 80, dtype: bool
df.duplicated(subset=['Survived', 'Pclass', 'Sex'])
False
False
9
       False
       True
10
14
       True
```

286 True299 True300 True

Length: 80, dtype: bool

COUNTING DUPLICATES AND NON DUPLICATES

df.Cabin.duplicated().sum() ==
11 df.duplicated().sum() 3

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 female		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 1	female	B77	S
	77	1	1	1 female B9		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 f	emale B9	S	
	134	1	2 f	emale	F33	S
	138	1	2 f	emale	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	(Cabin	Embarked	
0	0	1		male	C3	30 S	
1	1	1		female	D3	33 C	
9	1	3		male	E1	21 S	
10	1	1		female	B2	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	
299 300	1	1 2		female male	D2 F2		

77 rows x 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

import pandas as pd import numpy as np

74 rows x 5 columns

LABSHEET-3 DATA CLEANING

```
df=pd.read_csv('/content/2,1 dataset titanic.csv')
cols=['Name','Ticket','Cabin'] df=df.drop(cols,axis=1)
df.info()
Data columns (total 9 columns):
      # Column
                  Non-Null Count Dtype
         0 PassengerId
                         891 non-null int64
          1 Survived 891 non-null int64
          2 Pclass 891 non-null int64
          3 Sex 891 non-null object
4 Age 714 non-null float64
          4 Age
          5 SibSp 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
```

```
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  df=df.dropna() df.info()
  ₹<class 'pandas.core.frame.DataFrame'> Int64Index: 712 entries, 0 to 890
  Data columns (total 9 columns):
                      Non-Null Count Dtype
         # Column
                              712 non-null int64
             0 PassengerId
             1 Survived 712 non-null int64
             2 Pclass 712 non-null int64
                      712 non-null object
             3 Sex
             4 Age
                      712 non-null float64
             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:
        55.6+ KB
        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 1
                      0 22.0 1 0 7.2500 0 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
                                                  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
                                            30.0000 1 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
               9 9 1
        3 0
               a a 1
        4 1
               0 0 1
         885 0010
        886 1 0 0 1
        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()
  print(scaler.fit(data)) print('
                                             ')
  MinMaxScaler()
```

')

print(scaler.data_max_) print('
print('scaler.transform(data)')

☐ MinMaxScaler() [1. 18.]

20201ISB0027 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)

LABSHEET-4 Z-SCORE NORMALIZATION

```
import
          numpy
                          np
                                data=
                   as
[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)
\pm mean of the dataset ids 2.66666666666665 std is
3.3598941782277745
outlier in dataset is [15]
```

print('scaler.transform(data)')

MinMaxScaler() [1. 18.]
scaler.transform(data)

LABSHEET-5 OUTLIER DETECTION WITH IQR

Column

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()

₹ <class 'pandas.core.frame.DataFrame'> Index: 1436 entries, 0 to 1435 Data columns (total 10 columns): 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

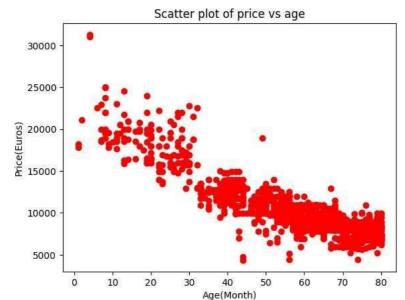
_													
₹		Price	Age	KM	FuelType	HP	MetC	olor	Autor	matic	cc	Doors	Weight
	0	13500 three		23.0 4698 1165	6.0	Dies	el	90		1.0	0		2000
	1	13750 3		23.0 7293 1165	7.0	Dies	el	90		1.0	0		2000
	3	14950 3		26.0 4800 1165	0.0	Dies	el	90		0.0	0		2000
	4	13750 3		30.0 3850 1170	0.0	Dies	el	90		0.0	0		2000
	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	Petr	ol	86		0.0	1		1300
	1424	7750		73.0 3471 1015	7.0	Petr	ol	86		0.0	0		1300
	5	1429 8950 1065		78.0 2400	0.0	Petr	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

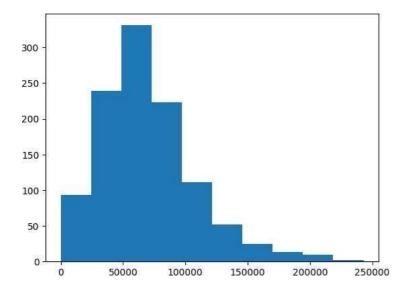
plt.scatter(df['Age'], df['Price'], c='red') plt.title('Scatter plot of price vs age') plt.xlabel('Age(Month)') plt.ylabel('Price(Euros)') plt.show()

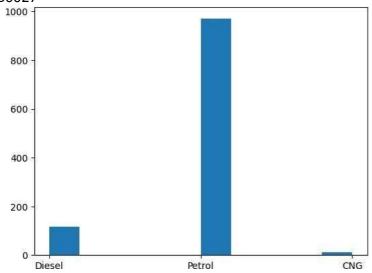


HISTOGRAM

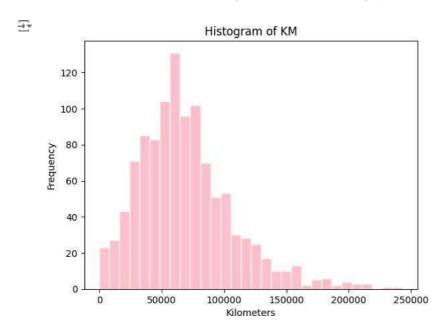
```
plt.hist(df['KM'])
```

```
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, 10., 20.]), array([1.0000000e+00, 2.430000e+04, 20.]), array([1.0000000e+00, 2.430000e+05]), array([1.00000000e+00, 2.430000e+05]), array([1.00000000e+00, 2.430000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.00000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.000000e+00, 2.430000e+00]), array([1.000000e+00, 2.430000e+00]), array([1.000000e+00, 2.430000e+00]), array([1.000000e+00, 2.430000e+00]), array([1.000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.430000e+00]), array([1.0000000e+00, 2.43000e+00]), ar
```



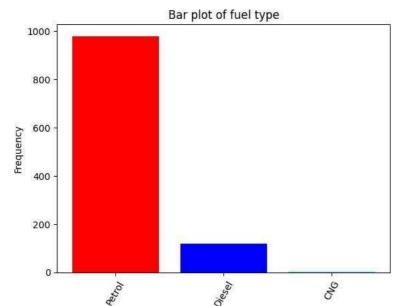


plt.hist(df['KM'],color='pink',edgecolor='white',bins=30) plt.title('Histogram of KM')
plt.xlabel('Kilometers') plt.ylabel('Frequency') plt.show()



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()

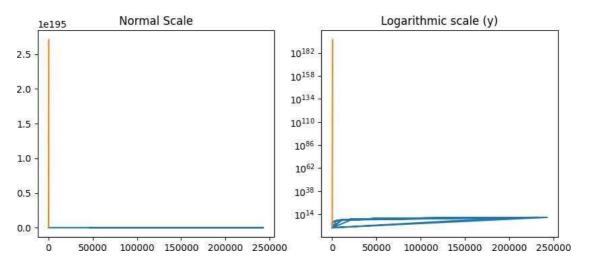


LINE PLOT

```
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)")
```

Fuel Types

/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)')



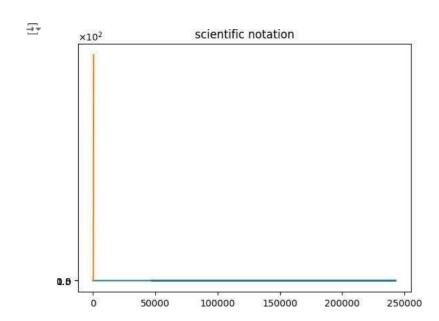
fig, ax = plt.subplots(figsize=(10,4)) x=df['KM']
ax.plot(x, x**2,x,x**3, lw=2) ax.set_xticks([1,2,3,4,5])
ax.set_xticklabels([r'\$/alphas',r'\$/beta\$',r'\$/gamma\$',r'\$/delta\$',
r'\$/epsilon\$'], fontsize=18) yticks=[0,50,100,150] ax.set_yticks(yticks)
ax.set_yticklabels(["\$%.1f\$" % y for y in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

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

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

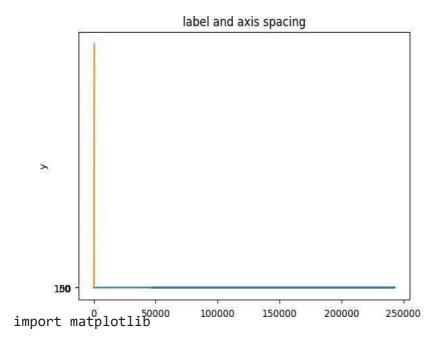
```
150.0 - The state of the state
```

```
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,
150])
        ax.set_title("label
                               and
                                       axis
                                                 spacing")
                  =
                              ax.yaxis.labelpad
ax.xaxis.labelpad
                         5
                                                   =
ax.set_ylabel("x") ax.set_ylabel("y") plt.show()
/usr/local/lib/python3.10/dist-
packages/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
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/')
'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
```

```
r.status_code = 200 url =
'https://www.romexchange.com/api?item=mastela&exact=false' headers =
{'UserAgent':'XY','Content-type':'application/json'}

r= requests.get(url, headers=headers)
r.status_code
= 500
r.text
= ...
```

LABSHEET-8 COLORMAPS

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

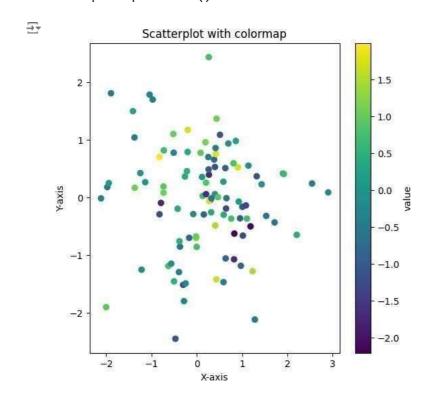
#sample datafame with multiple columns

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()



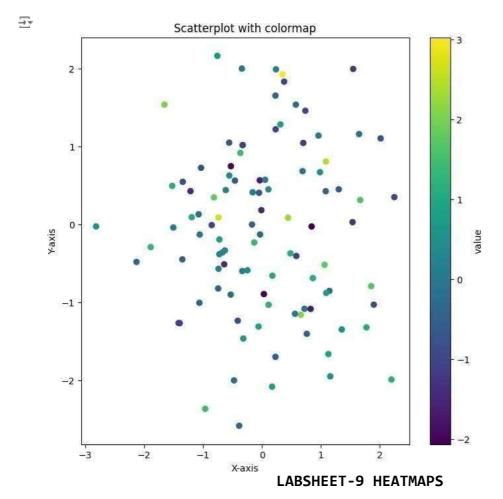
import pandas as pd import numpy as np
print(np.random.randn(100))

[-1.65970274e-01 -3.26301492e-01 -6.97091694e-01 5.29185683e-01 1.65900203e-01 2.57310809e-01 1.87945887e-01 -1.47856355e+00 1.85465880e+00 -5.74773399e-02 -7.28047219e-01 1.43513290e+00 1.16276640e-01 3.62925427e-01 2.27296732e-01 -4.68725785e-01 -7.20465601e-01 2.31190101e-01 5.47647007e-01 6.14310198e-01 -2.88178116e-01 -2.59650445e-01 7.14726089e-02 2.91407763e-01 7.44199514e-01 1.03744520e+00 5.19583750e-02 -1.22315192e+00 2.82553552e-01 9.27484581e-01 4.68496647e-01 3.97669795e-01 -6.15495640e-01 -3.59199216e-01 1.45247374e-01 -1.61267440e-01 -1.08796055e+00 2.03942727e-01 1.33177945e-03 7.08911052e-01 1.92045492e+00 1.06460553e+00 9.71054014e-01 8.14301945e-01 1.01645092e-01 -9.38076692e-02 1.33631841e+00 2.55274328e-01

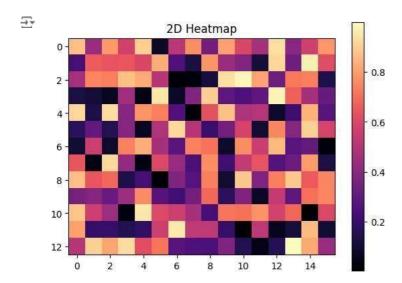
-5.17379367e-01 -1.71773916e+00 9.24194703e-01 1.67657214e-01

```
-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
```

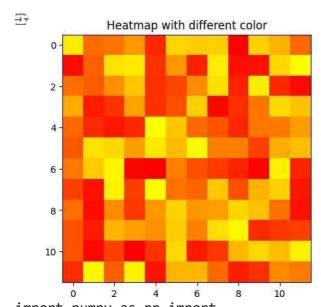
#sample datafame with multiple columns
data=pd.DataFrame({"x":np.random.randn(100),"y":np.random.randn(100),"value":np.random.
ran dn(100)}) #define the colormap and alpha values cmap="viridis" alpha=1 #create the
scatterplot plt.figure(figsize=(8,8))
plt.scatter(data["x"],data["y"],c=data["value"],cmap=cmap,alpha=alpha) #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()



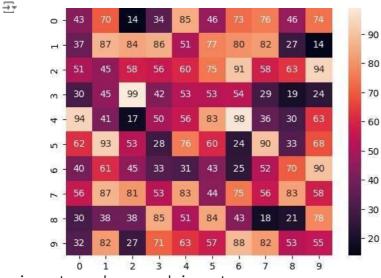
import matplotlib.pyplot as plt data= np.random.random((13,16))
plt.imshow(data,cmap="magma") plt.title("2D Heatmap") plt.colorbar()
plt.show()



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()



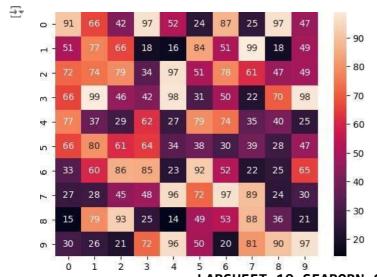
import numpy as np import
seaborn as sns import
matplotlib.pyplot as plt
data= np.random.randint(low=14,high=100, size=(10,10))
hm=sns.heatmap(data=data, annot=True) plt.show()



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

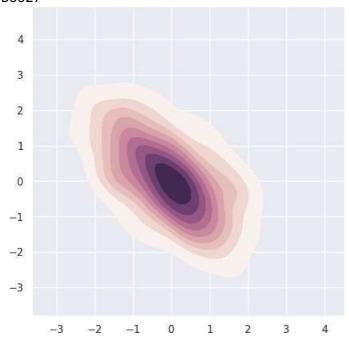
import numpy as np import pandas as pd
import matplotlib.pyplot as plt import seaborn as sns
%matplotlib inline

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

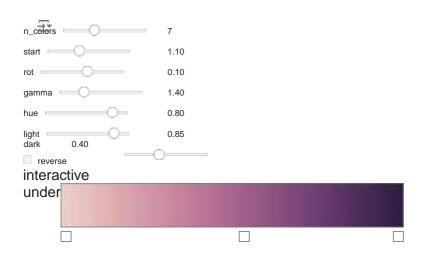
BUILDING COLOR PALLETTES

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

<Axes: >



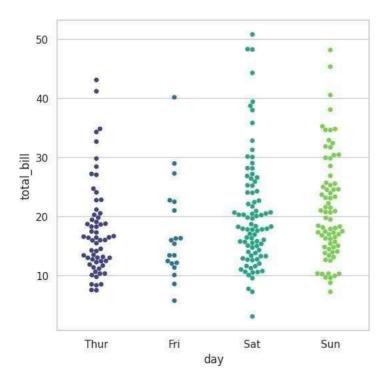
sns.choose_cubehelix_palette(as_cmap=True)



sns.set_style('whitegrid')

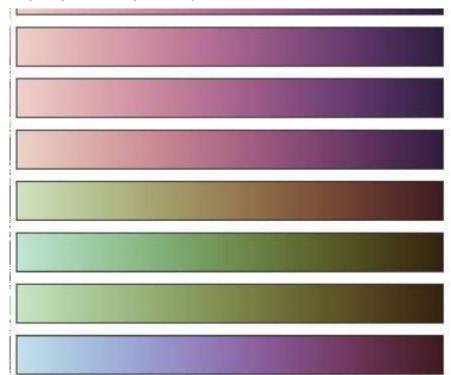
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l sns.swarmplot(x="day", y="total_bill", data=tips, palette="viridis")

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

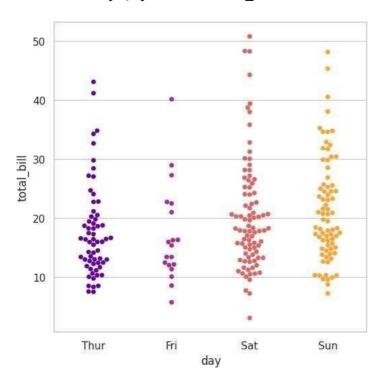


sns.set_style('whitegrid')

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")

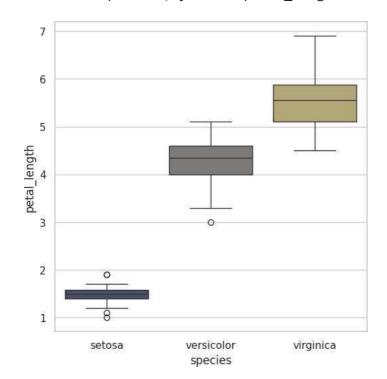


<Axes: xlabel='day', ylabel='total_bill'>



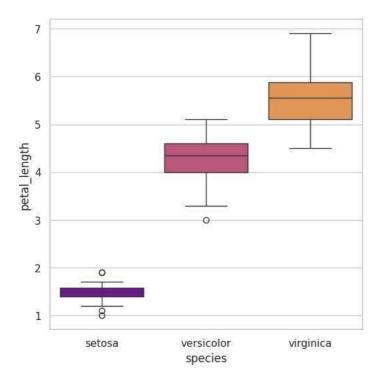
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l sns.boxplot(x="species", y="petal_length", data=iris, palette="cividis")

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



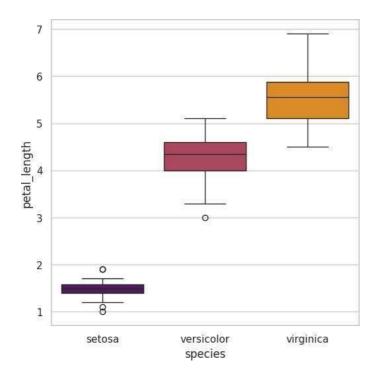
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l sns.boxplot(x="species", y="petal_length", data=iris, palette="plasma")

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



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'>



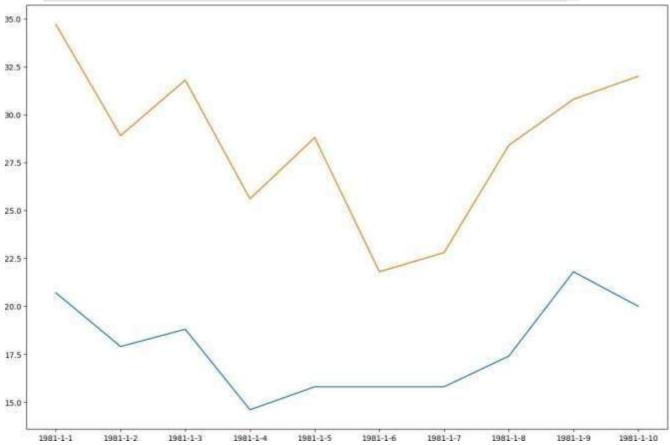
LABSHEET-11 MULTIVARIATE VISUALIZATION

```
Relational plots: relation b/w two variables categorical plots: categorical values are
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
                         axes.plot(dates,min_temperature,
figsize=(15,10))
label='Min temperature')
```

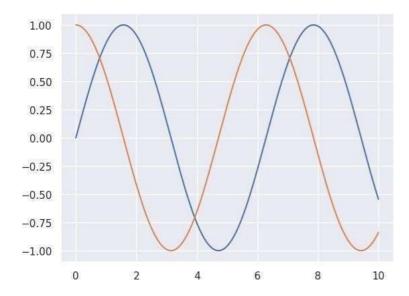
```
axes.plot(dates,max_temperature, label='Max temperature') axes.legend
```

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

Call signatures::



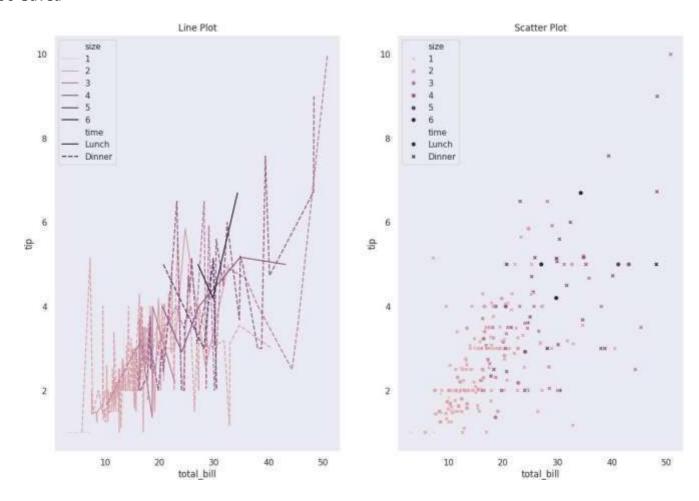
matplotlib.axes._axes.Axes.legend sns.set()



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")
        = sns.scatterplot(x="total_bill", y="tip",
                                                               hue="size", style="time",
sct_plt
                                ax=ax[1]).set_title("Scatter
data=df,
                                                                                       Plot")
sct_plt.figure.savefig('Scatter_plot1.png') print('Plot Saved')
          total_bill tip sex smoker day time size
                   No Sun Dinner
16.99 1.01 Female
10.34 1.66
             Male
                   No Sun Dinner
                                 3
                   No Sun Dinner
21.01 3.50
            Male
                                 3
23.68 3.31
             Male
                   No Sun Dinner
                                 2
24.59 3.61 Female
                   No Sun Dinner
```

Plot Saved



```
20201ISB0027
 df=sns.load_dataset('tips')
 sns.barplot(x='sex', y='total_bill', data=df, palette='plasma', estimator=
 np.std, ax=ax[0,0]).set title('Bar Plot') sns.countplot(x='sex',
                                                                       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",

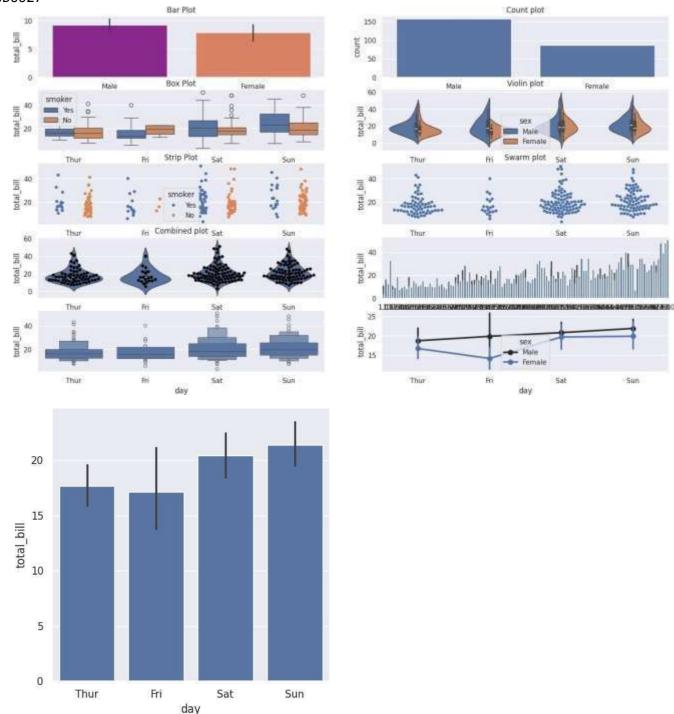
79e72dcff921>:26: FutureWarning:

hue="sex", data=df, ax=ax[4,1])

<seaborn.axisgrid.FacetGrid at 0x7e3ac3b802e0>

y="total_bill", color="b", scale="linear", data=df, ax=ax[4,0]) <ipython-input-6-

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",



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

0     5.1 3.5 1.4 0.2 setosa

1     4.9 3.0 1.4 0.2 setosa

2     4.7 3.2 1.3 0.2 setosa

3     4.6 3.1 1.5 0.2 setosa

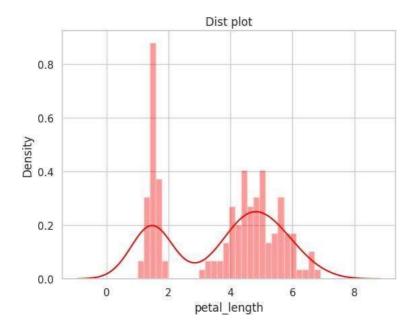
4     5.0 3.6 1.4 0.2 setosa
```

`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')



`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).

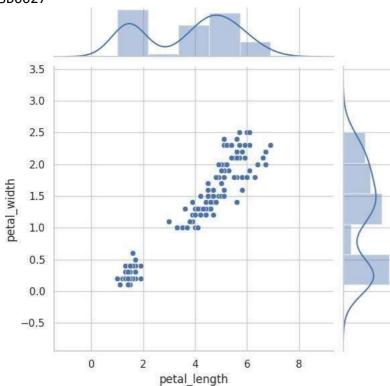
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751 func(self.x, **orient_kw_x, **kwargs)

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

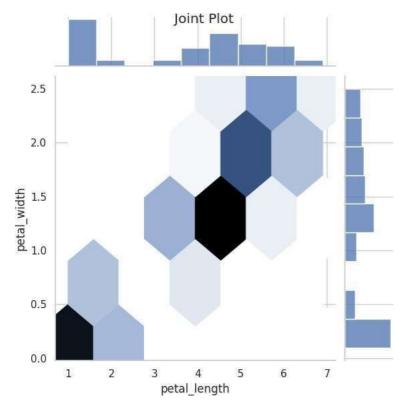
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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')



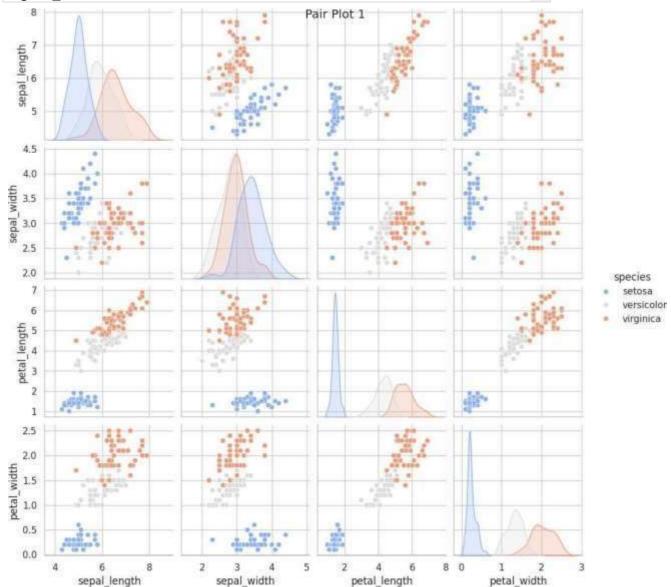
g=sns.pairplot(df, hue="species", palette= 'coolwarm') g.fig.suptitle("Pair Plot 1")
g.add_legend

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.

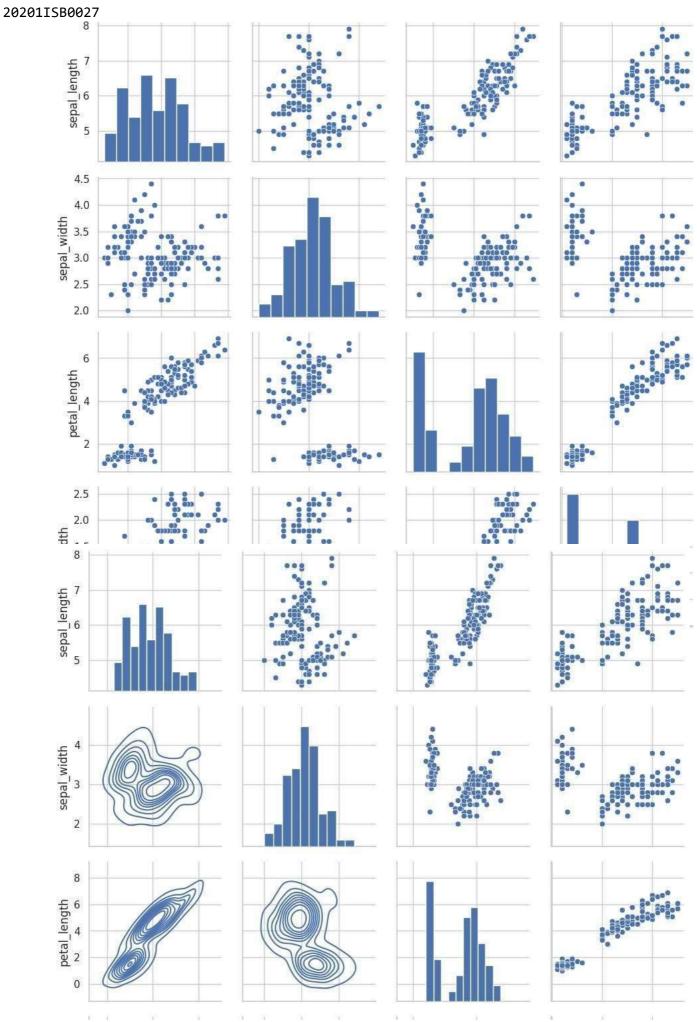
Parameters

legend_data : dict



```
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()





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 th 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 Next steps: Generate code with df View recommendedplots

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'.

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

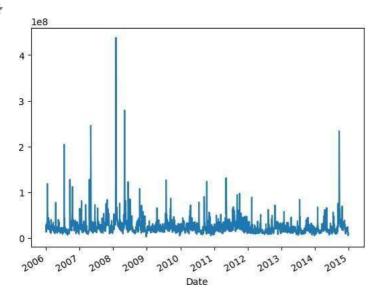
```
Open High Low Close Volume

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
```

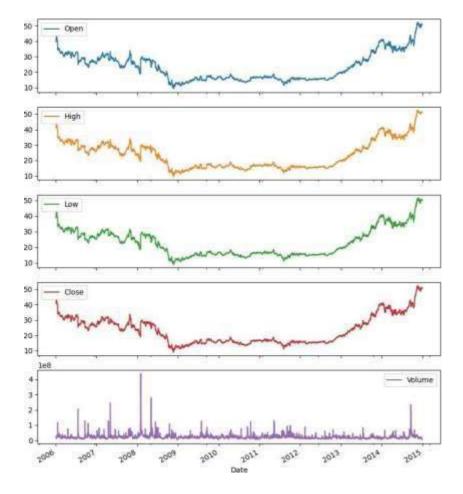
[2263 rows $x \ 5 \ columns$]

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.



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:

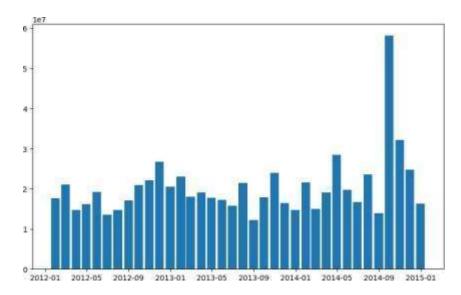
```
# Resampling the time series data based on monthly 'M' frequency df_month
= df.resample("M").mean() print(df_month)

# 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')
```

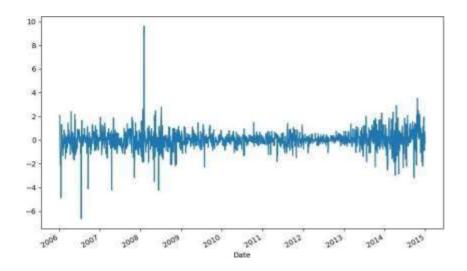
Ŧ		0pe	n Hi	gh	Low C	lose	Volume	Change
	Date							
	2006-01-31	38.245500	38.694000	37.641500	38.113000	3.400594e+0	7 0.991442	
	2006-02-28	33.141579	33.436842	32.627368	32.975789	2.329848e+0	7 0.996423	
	2006-03-31	31.333478	31.696957	30.929130	31.218696	2.095522e+0	7 1.000390)
	2006-04-30	32.383684	32.790000	31.914737	32.283158	2.200768e+0	7 1.001098	
	2006-05-31	31.744545	32.175455	31.171364	31.517273	2.218047e+0	7 0.998535	
	2014-08-31	36.836190	37.150000	36.545238	36.876667	1.396539e+0	7 1.003530)
	2014-09-30	40.662857	41.270000	39.983810	40.671905	5.811769e+0	7 1.003005	
	2014-10-31	41.253043	41.886087	40.784783	41.393913	3.210848e+0	7 1.005501	
	2014-11-30	49.879474	50.553158	49.440000	50.151579	2.474402e+0	7 1.006233	
	2014-12-31	50.359524	50.975714	49.852857	50.331905	1.623090e+0	7 0.999653	

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



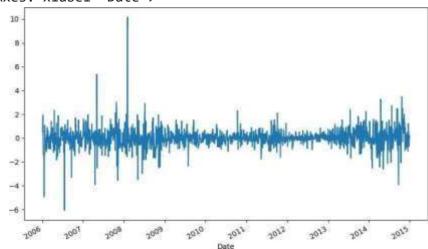
Differencing: Differencing is used to make the difference in values of a specified interval. By default, it's one, we can specify different values for plots. It is the most popular method to remove trends in the data.

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



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





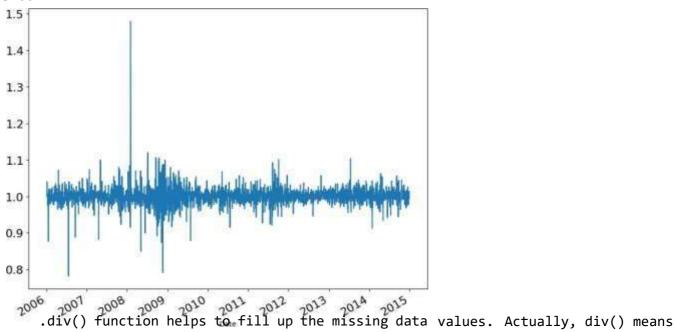
Plotting the Changes in Data

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))

