# DATA HANDLING AND VISUALIZATION LABSHEETS

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**COURSE CODE: CSE2026** 

# LABSHEET-1 INTRODUCTION TO NUMPY

```
import numpy as np a=np.array([1,2,3])
 b=np.array([1,2,3])
add=np.add(a,b)
 \pm array([2, 4, 6])
a=np.array([5,10,20])
 b=np.array([4,8,10])
sub=np.subtract(a,b)
                       sub
 \pm array([ 1, 2, 10])
a=np.array([5,10,20])
 b=np.array([4,8,10])
sub=np.multiply(a,b)
                       sub
 a=np.array([5,7,9]
 b=np.array([4,5,6]
 )
sub=np.mod(a,b)
 \pm array([1, 2, 3])
a=np.array([1,2,3])
 b=np.array([1,2,3])
 add=np.power(a,b) add
 \pm 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
 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
          а
      102
      103
          d dtype:
      104
```

```
Series with Dictionary
import pandas as pd import numpy as np
 data={'a':
               0.,'b':
                          1.,'c':
 s=pd.Series(data) print(s)
 b 1.0 c
 2.0
 dtype: float64
 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)
글 • b
      1.0 c 2.0
      d NaN a
      0.0
 dtype: float64
 Create Series from Scalar
import pandas as pd import numpy as np
 s= pd.Series(5, index=[0,1,2,3])
 print(s)
   ₹ 0
        ₅ dtype:
      3
 int64
 Retrieving data from the zeroth position
 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]) ** a
 100 b
      101 c
 102
 dtype: int64
import pandas as pd
```

```
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  s=
  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
       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 s=
  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[['a','e','i','d']])
≟ a
       100 e
       104 i
       108 d 103
  dtype: int64
  Data Frames
 import pandas as pd
  df=pd.read_csv("/content/nyc_weather.csv")
  Create data frame with empty data
 import pandas as pd df=pd.DataFrame()
  print(df)
  Index: []
  Create data frame from list
  import
            pandas
                              pd data=[1,2,3,4,5]
                        as
  df=pd.DataFrame(data) print(df)
 ₹
  0
    1
  1
    2
```

```
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  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
 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)
          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
       Ricky 29
                                          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
                   male
                         C30
                               S 1
      1
            1
                   female D33
                               С
          1 3
                         E121
                   male
   10
          1 1
                   female B22
                               S
```

**14** 0

male

B51 B53 B55 S

```
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FINDING DUPLICATE ROWS

df.Cabin.duplicated()

False

False
```

```
df.Cabin.duplicated()
 9
       False
       False
 10
 14
       False
 . . .
       False
 271
       False
 278
       False
 286
 False
 False
 Name: Cabin, Length: 80, dtype: bool
df.duplicated()
 False
 False
 9
       False
 10
       False
       False
 14
       False
 271
 278
       False
       False
 286
 False
 False
 Length: 80, dtype: bool
df.duplicated(subset=['Survived', 'Pclass', 'Sex'])
 False
 False
 9
       False
 10
       True
       True
 14
       True
 271
 278
       True
 286
        True
 299
        True
 300
        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(), :]
   Ŧ
           Survived Pclass
                             Cabin Embarked
                         Sex
       138
                  2 female F33
                  1 female B77
```

**237** 1 1 female B96 B98 S

# **USING KEEP**

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

Ŧ		Survived	Pclass	Sex	Cabin	Embarked
	138	1	2 female	F33	S	
	169	1	1 female	B77	S	
	237	1	1 female E	396 B98	S	

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

<del>_</del>		Survived	Pclass	Sex	Cabin	Embarked	
	36	1	1	female	B77	S	
	77	1	1	S			
	134	1	2	female	F33	S	

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

<del>-</del>						
		Survived	Pclass	Sex	Cabin	Embarked
	36	1	1 fe	emale	B77	S
	77	1	1 fe	emale B9	96 B98	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 fe	emale B9	96 B98	S

# DROPPING DUPLICATED ROWS

df.drop\_duplicates()

→ ÷						
_		Survived	Pclass	Sex	Cabin	Embarked
	0	0	1	male	e C	30 S
	1	1	1	fema	ale D33 m	nale C
	9	1	3	E12 <sup>2</sup>	1 female E	322 S
	10	1	1	male B51 E	853 B55	S
	14	0	1		male	S
					C93 male	
	271	1	1	C111 male	5148	S
	278	0	1			С
	286	1	1			С
	299 300	1	1 2	fema male		s s

77 rows × 5 columns

df.drop\_duplicates(keep=False)

→ ×							
		Survived	Pclass	Sex		Cabin	Embarked
	0	0	1		male	C30	S
	1	1	11	emale	D33		С
	9	1	3		male	E121	S
	10	1	11	emale	B22		S
	14	0	1		male B5	1 B53 B	55 S
	271	1	1		male	C93	S
	278	0	1		male	C111	C
	286	1	1		male	C148	3 C
	299	1	11	emale	D21		S
	300	1	2		male	F2	S

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):
                    Non-Null Count Dtype
         # Column
             0 PassengerId
                           891 non-null int64
             1 Survived 891 non-null int64
             2 Pclass 891 non-null int64
                    891 non-null object
             3 Sex
             4 Age
                      714 non-null float64
             5 SibSp 891 non-null int64
             6 Parch 891 non-null int64
                     891 non-null float64
               8 Embarked 889 non-null object dtypes: float64(2), int64(5), object(2) memory usage:
        .8+ KB
  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
                      712 non-null object
             3 Sex
                     712 non-null float64
             5 SibSp
                      712 non-null int64
             6 Parch 712 non-null int64
                     712 non-null float64
             8 Embarked 712 non-null object dtypes: float64(2), int64(5), object(2) memory usage:
        55.6+ KB dummies=[] cols=['Pclass','Sex','Embarked'] for col in cols:
        dummies.append(pd.get_dummies(df[col]))
```

```
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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 22.0 1 0 7.2500 0 0 1
            1 2
                 1 38.0 1 0 71.2833 1 0 0 1
                1 26.0 0 0 7.9250 0 0 1
            2 3
            3 4
                 1 35.0 1 0 53.1000 1 0 0 1
            886
            887
                       27.0 0
                                         13.0000 0 1 0
                  0
                        19.0 0
                                    0
       887
            888
                 1 15.0
1 889 890 1
                  1
                                         30.0000 1 0 0
                                9 30.0000 1 0
26.0 0 0
                 0 32.0 0
                                  0 7.7500 0 0 1
       890
            891
       male C Q S 0 1 0
          1
                  100
           0
                 001
                 001
                 001.. .....
       885
          0010
          1001
       886
       887 0001
       889 1100
       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)')
  <sup>™</sup>MinMaxScaler() [ 1. 18.]
  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]
```

```
[[ 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
                                   data=
           numpy
                       as
  [1,2,2,2,3,1,1,15,2,2,2,3,1,1,2]
  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)
 oxdot mean of the dataset ids 2.6666666666666665 std is
  3.3598941782277745 outlier
  in dataset is [15]
  LABSHEET-5 OUTLIER
  DETECTION WITH IQR
  import numpy as np import
  seaborn as sns
     data=[6,2,3,4,5,1,50]
  sort_data=np.sort(data) sort_data
\pm array([ 1, 2, 3, 4, 5, 6, 50])
  Q1=-np.percentile(data, 25, interpolation = 'midpoint') Q2=-np.percentile(data, 50,
  interpolation = 'midpoint') Q3=-np.percentile(data, 75, interpolation = 'midpoint')
 print('Q1 25 percentile of the given data is, ', Q1) print('Q2 50 percentile of the
  given data is, ', Q2) print('Q3 75 percentile of the given data is, ', Q3)
  IQR = Q3 - Q1 print('IQR is', IQR)
  ₹ Q1 25 percentile of the given data is, -2.5 Q2 50 percentile of the given data is,
  4.0 Q3 75 percentile of the given data is, -5.5 IQR is -3.0 low_lim
  = Q1 - 1.5 * IQR up_lim = Q3 + 1.5 * IQR
```

[8.8e+01 7.0e-02] [4.0e+00 1.0e-01]]

# 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

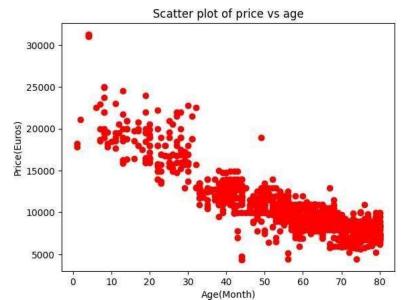
	Price Ag	ge KM	FuelType	HP	MetColor	Automatic	CC	Doors	Weight
0	13500 three	23.0 46986.0 1165	] (	Diesel	90	1.0	0	20	000
1	13750 23.0	72937.0 Dies	el 90 1.0 0	2000 3	1165				
3	14950 26.0	48000.0 Dies	el 90 0.0 0	2000 3	1165				
4	13750 30.0	38500.0 Dies	el 90 0.0 0	2000 3	1170				
5	12950 3	32.0 61000.0 1170	) [	Diesel	90	0.0	0	20	000
1423	7950 3	80.0 35821.0 1015	) [	Petrol	86	0.0	1	13	300
1424	7750 3	73.0 34717.0 1015	) F	Petrol	86	0.0	0	13	300
5	<b>1429</b> 8950 1065	78.0 24000.0	) F	Petrol	86	1.0	1	13	300

**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

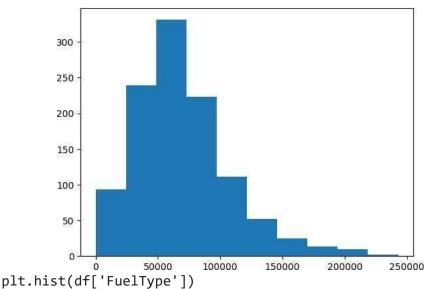
# **SCATTER PLOT**

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()

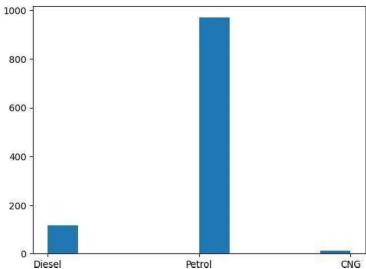


# **HISTOGRAM**

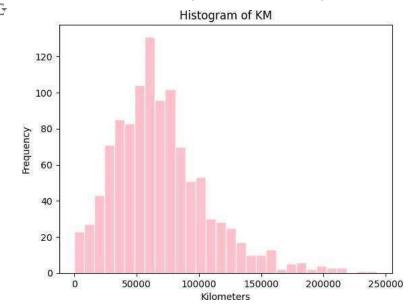
```
plt.hist(df['KM'])
```



(array([117., 0., 0., 0., 970., 0., 0., 12.]), array([0. , 0.2, 0.4, 0.6, 0.8, 1. , 1.2, 1.4, 1.6, 1.8, 2. ]), <BarContainer object of 10 artists>)

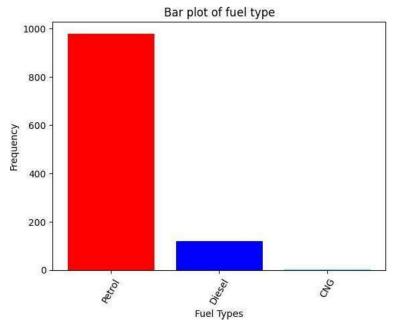


plt.hist(df['KM'],color='pink',edgecolor='white',bins=30) plt.title('Histogram of KM')
plt.xlabel('Kilometers') plt.ylabel('Frequency') plt.show()



# **BAR PLOT**

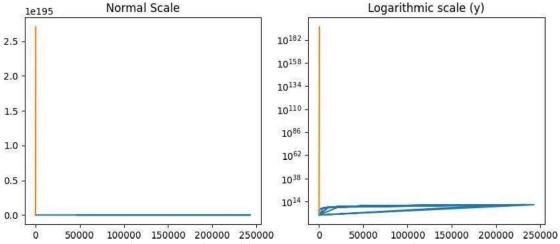
```
counts = [979,120,2] fueltype= ('Petrol','Diesel','CNG') index=
np.arange(len(fueltype))
plt.bar(index,counts,color=['red','blue','cyan']) plt.title('Bar plot of fuel type')
plt.xlabel('Fuel Types') plt.ylabel('Frequency') plt.xticks(index, fueltype,
rotation= 60) plt.show()
```



#### 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)")
```

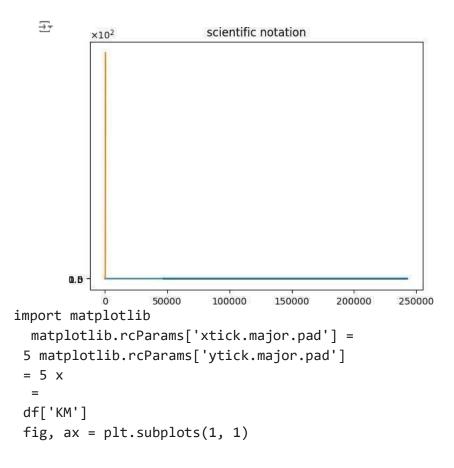
/usr/local/lib/python3.10/dist-packages/pandas/core/arraylike.py:396:
RuntimeWarning: overflow encountered in exp result = getattr(ufunc, method)(\*inputs,
\*\*kwargs) Text(0.5, 1.0, 'Logarithmic scale (y)')



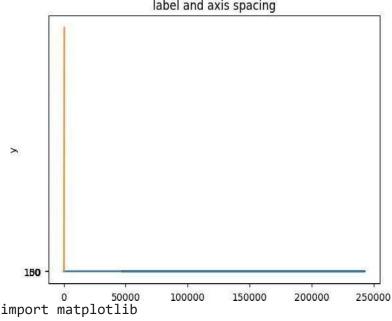
fig, ax = plt.subplots(figsize=(10,4)) x=df['KM'] ax.plot(x, x\*\*2,x,x\*\*3,
lw=2) ax.set\_xticks([1,2,3,4,5])
ax.set\_xticklabels([r'\$/alphas',r'\$/beta\$',r'\$/gamma\$',r'\$/delta\$',
r'\$/epsilon\$'], fontsize=18) yticks=[0,50,100,150] ax.set\_yticks(yticks)
ax.set\_yticklabels(["\$%.1f\$" % y for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

```
[Text(0, 0, '$0.0$'),
Text(0, 50, '$50.0$'),
Text(0, 100, '$100.0$'),
Text(0, 150, '$150.0$')]
```

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



```
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  ax.plot(x, x^{**2}, x, np.exp(x)) ax.set_yticks([0, 50, 100, 150])
  ax.set_title("label and axis spacing")
  ax.xaxis.labelpad
                                  ax.yaxis.labelpad
                                                             5 ax.set ylabel("x")
                             5
                                                       =
  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)
                        label and axis spacing
```



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

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

headers = {'Content-type': 'application/json'}

```
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/')

Crested wheel for library Successfully installed library-0.0.0
```

```
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ur l

    '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/'

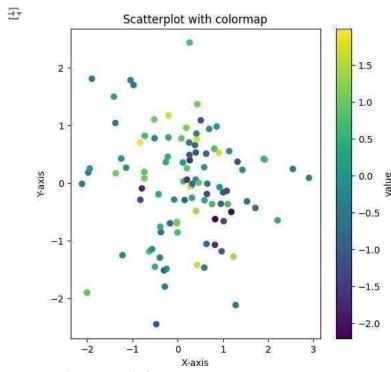
 header
S
 r
 r.status_code = 200 url =
  'https://www.romexchange.com/api?item=mastela&exact=false' headers =
 {'UserAgent':'XY','Content-type':'application/json'}
r= requests.get(url, headers=headers)
 r.status_code
 ∓ 500
```

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))

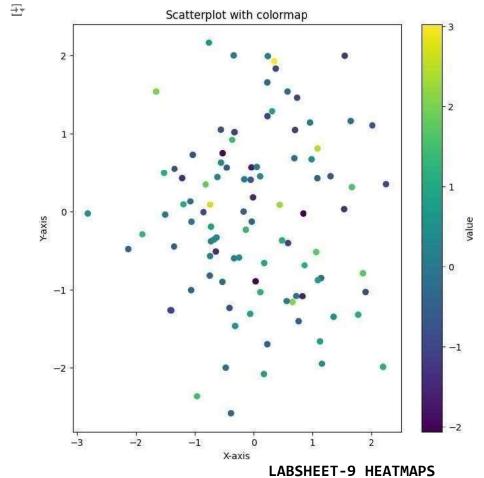
- $\pm$  [-1.65970274e-01 -3.26301492e-01 -6.97091694e-01 5.29185683e-01 1.65900203e-01
- 2.57310809e-01 1.87945887e-01 -1.47856355e+00
- 1.85465880e+00 -5.74773399e-02 -7.28047219e-01 1.43513290e+00 1.16276640e-
- 01 3.62925427e-01 2.27296732e-01 -4.68725785e-01
- -7.20465601e-01 2.31190101e-01 5.47647007e-01 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

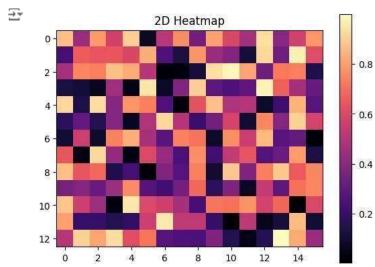
matplotlib.pyplot as plt

#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 numpy as np
import
matplotlib.pyplot as

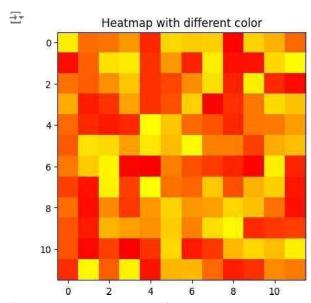
plt data=
np.random.random((13
,16)) plt.imshow(
data,cmap="magma")
plt.title("2D
Heatmap")
plt.colorbar()
plt.show()



import numpy as np

import matplotlib.pyplot as plt data=np.random.random((12,12)) plt.imshow(data,
cmap='autumn')

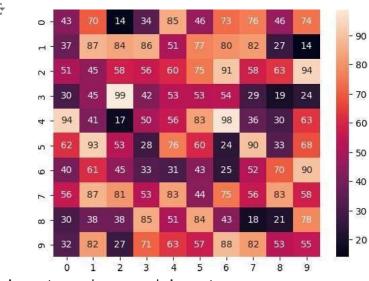
plt.title("Heatmap with different color") plt.show()



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()



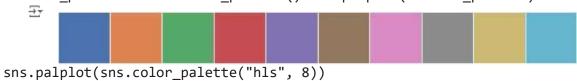
LABSHEET-10 SEABORN COLOR PALLETTES

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

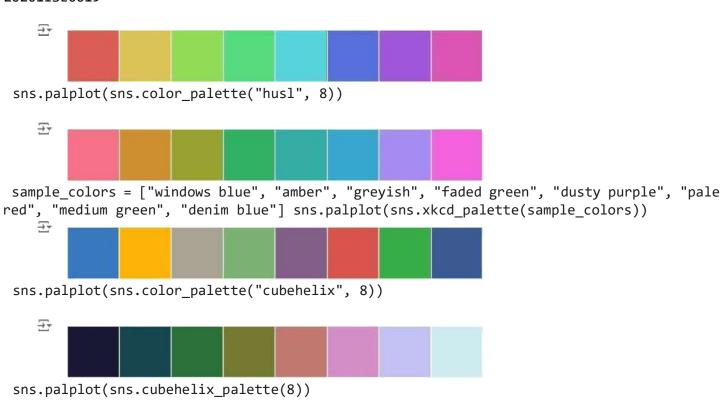
sns.set(rc={"figure.figsize": (6,6)})

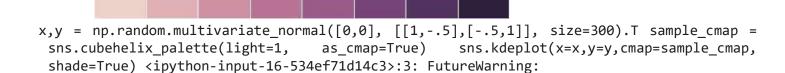
# **BUILDING COLOR PALLETTES**

current\_palette = sns.color\_palette() sns.palplot(current\_palette)



₹



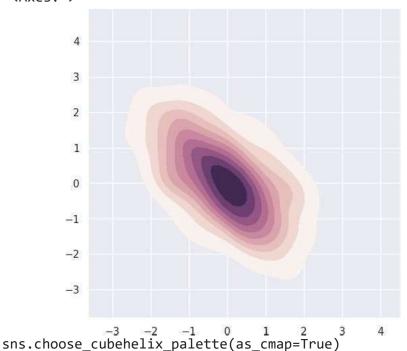


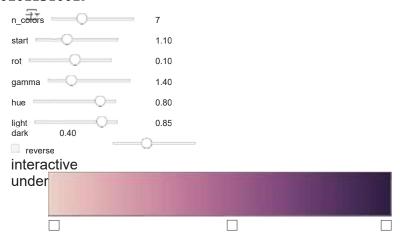
`shade` is now deprecated in favor of `fill`; setting `fill=True`.

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: >

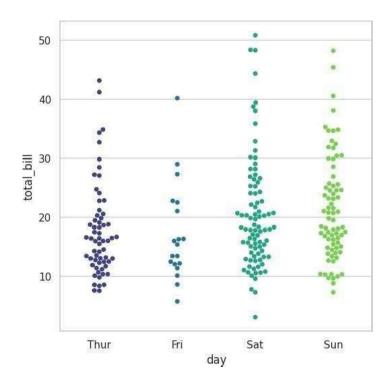




sns.set\_style('whitegrid')

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")

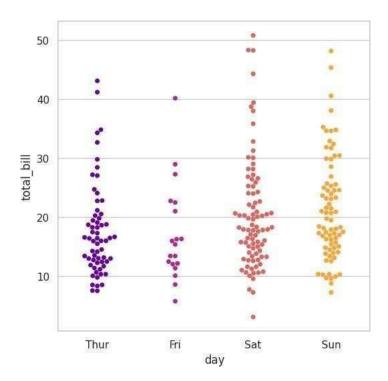
<Axes: xlabel='day', ylabel='total\_bill'>



sns.set\_style('whitegrid') sns.swarmplot(x="day", y="total\_bill", data=tips,
palette="plasma") = <ipython-input-24- 8931cda8de8a>: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")

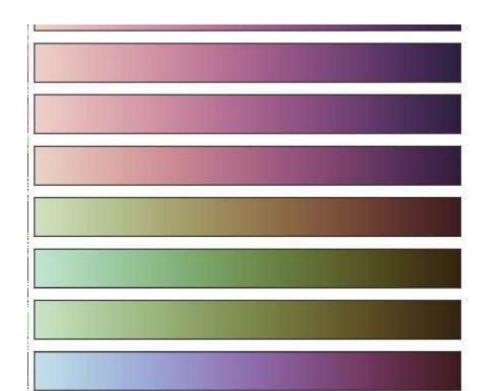
<Axes: xlabel='day', ylabel='total\_bill'>

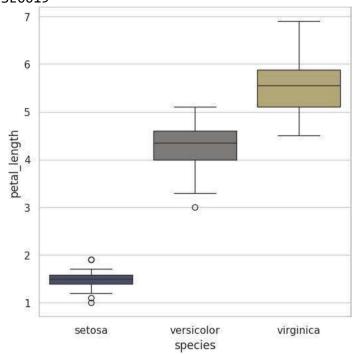


iris = sns.load\_dataset("iris")

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'>

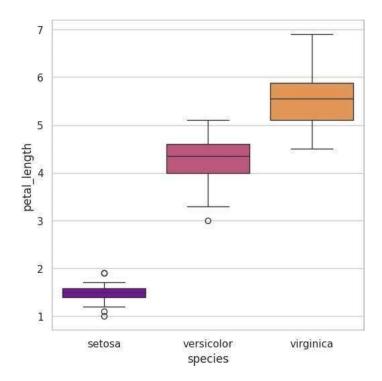




iris = sns.load\_dataset("iris") sns.boxplot(x="species", y="petal\_length", data=iris,
palette="plasma") = <ipython- input-27-0b4fe890c1f3>:2: FutureWarning:

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.boxplot(x="species", y="petal\_length", data=iris, palette="plasma")

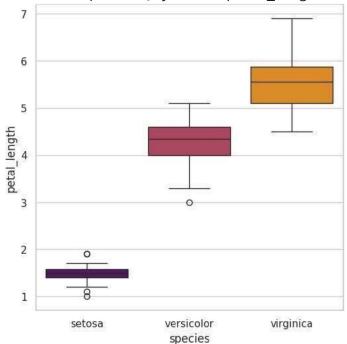
<Axes: xlabel='species', ylabel='petal\_length'>



iris = sns.load\_dataset("iris")

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'>



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="magma")

<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]

fig,axes = plt.subplots(nrows=1, ncols=1, figsize=(15,10))

axes.plot(dates,min\_temperature, label='Min temperature')

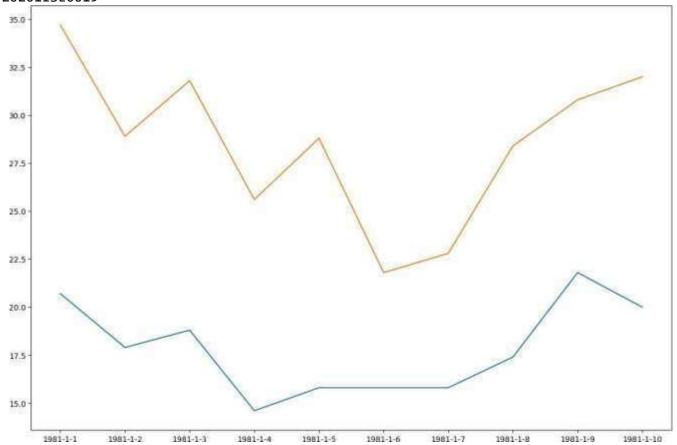
axes.plot(dates,max temperature, label='Max temperature') axes.legend

```
def legend(*args,
  **kwargs)

/usr/local/lib/python3.10/dist-
packages/matplotlib/axes/_axes.py

Place a legend on the Axes.

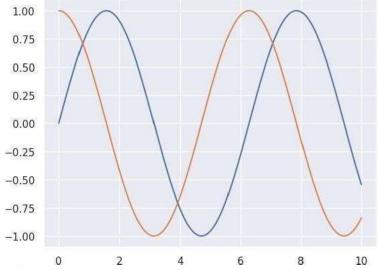
Call signatures::
```



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

```
x = np.linspace(0,10,1000)
plt.plot(x, np.sin(x), x, np.cos(x))

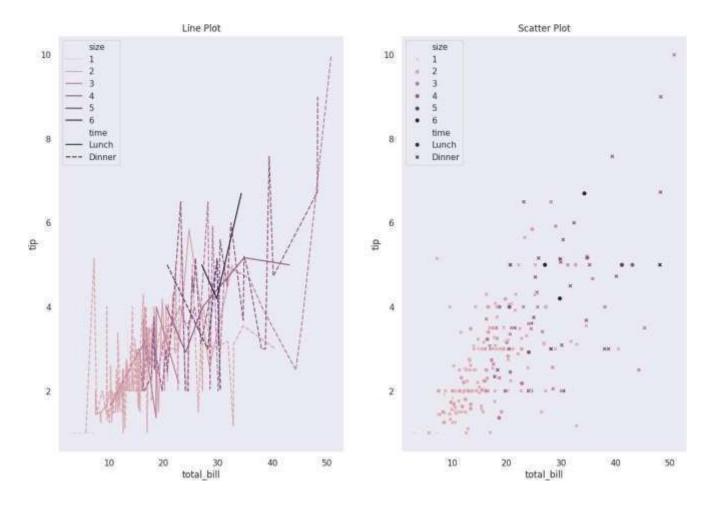
  [<matplotlib.lines.Line2D at 0x7e3acaaaffa0>,
<matplotlib.lines.Line2D at 0x7e3acaae0040>]
```



sns.set(style="dark")
fig, ax =
plt.subplots(ncols=2,
nrows=1,
figsize=(15,10)) df=
sns.load\_dataset("tip
s") 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", data=df,
ax=ax[1]).set_title("Scatter Plot")
sct_plt.figure.savefig('Scatter_plot1.png') print('Plot Saved') =
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
                    No Sun Dinner
             Male
             Male
                    No Sun Dinner
23.68 3.31
                                  2
24.59 3.61 Female
                    No Sun Dinner
```

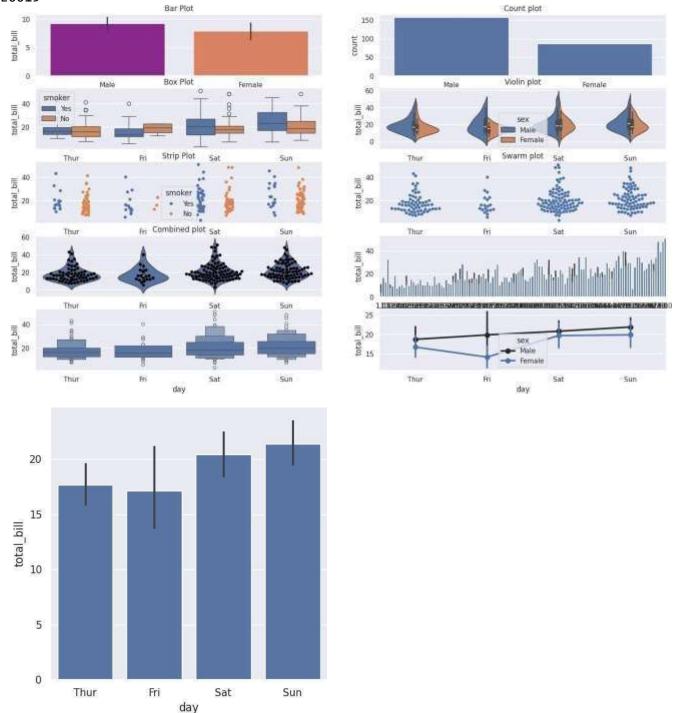
#### Plot Saved



```
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')
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,
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]) sns.catplot(x='day',y='total_bill',data=df, kind='bar') \(\frac{1}{27}\) <ipython-input-
6- 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- 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>

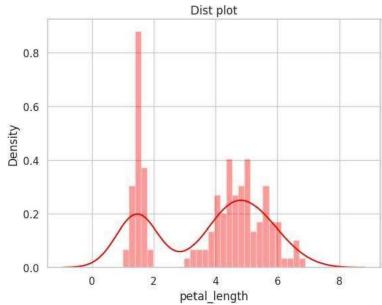


sns.set\_style('whitegrid')
#loading the dataset directly without any files df=sns.load\_dataset('iris')
print(df.head())

`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 <a href="https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751">https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751</a> 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 <a href="https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751">https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751</a> func(self.x, \*\*orient\_kw\_x,

\*\*kwargs)

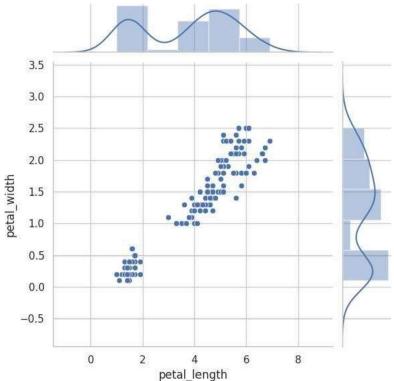
/usr/local/lib/python3.10/dist-packages/seaborn/axisgrid.py:1892: UserWarning:

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

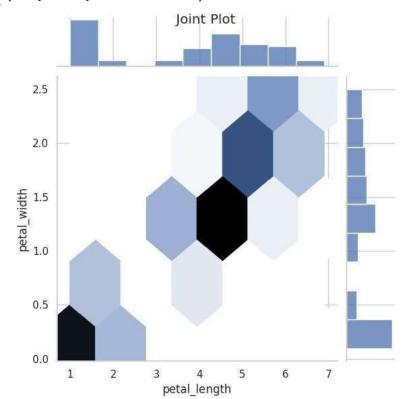
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms). For a guide to updating your code to use the new functions, please see <a href="https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751">https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751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

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')



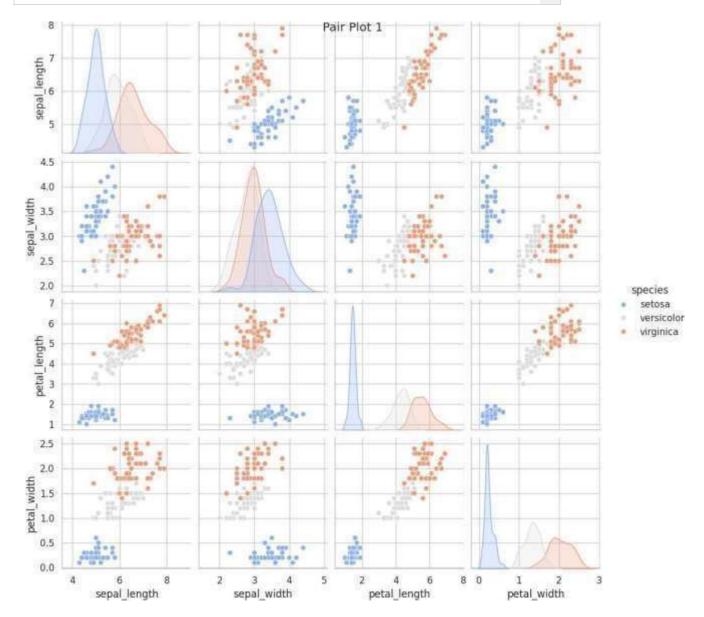
g=sns.pairplot(df, hue="species", palette= 'coolwarm') g.fig.suptitle("Pair Plot 1")
g.add\_legend

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

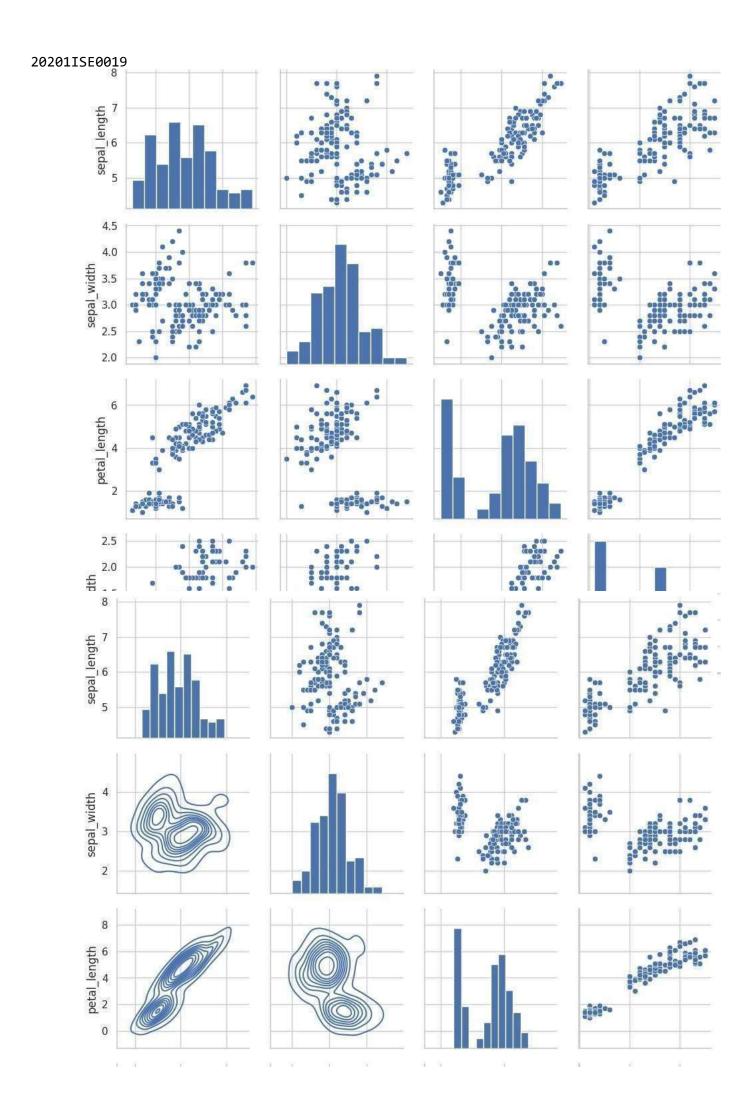
Draw a legend, maybe placing it outside axes and resizing the figure.

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)
```

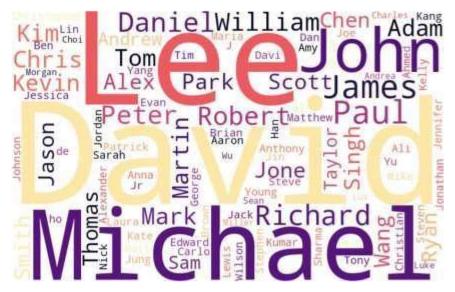


# LABSHEET-12 TEXT VISUALIZATION

```
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...
 Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...
 ndf=df.dropna() ndf.head()
   Ŧ
 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(background_color="White").generate(text) plt.imshow(wordcloud,
 interpolation= 'bilinear') plt.axis("off")
 plt.margins(x=0, y=0) plt.show()
                 Alex Richard
       Daniel
         Chris
                 Scott
 wordcloud = WordCloud(background_color="White", max_words=100, max_font_size=300, width=
 800,
                             colormap="magma").generate(te
           height=500,
 plt.figure(figsize=(20,20)) plt.imshow(wordcloud,
```

'bilinear') plt.axis("off") plt.margins(x=0, y=0) plt.show()





Next steps:

# 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

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'.

Viewrecommendedplots

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

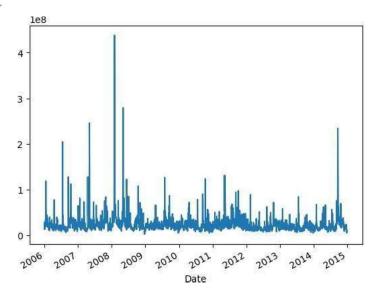
Generate code with d

```
Ŧ
                 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]

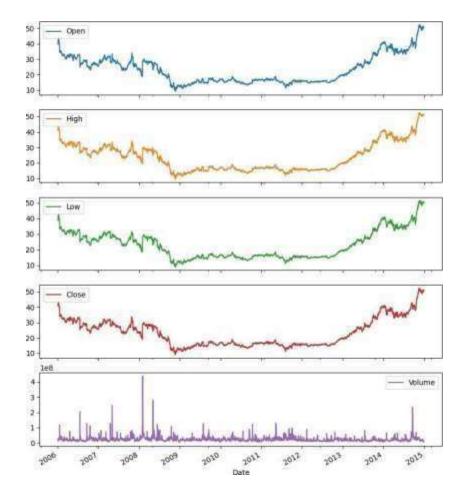
Example 1: Plotting a simple line plot for time series data.

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



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

```
df.plot(subplots=True, figsize=(10, 12))
    array([<Axes: xlabel='Date'>, <Axes:
xlabel='Date'>,
    <Axes: xlabel='Date'>, <Axes: xlabel='Date'>,
    <Axes: xlabel='Date'>], dtype=object)
```



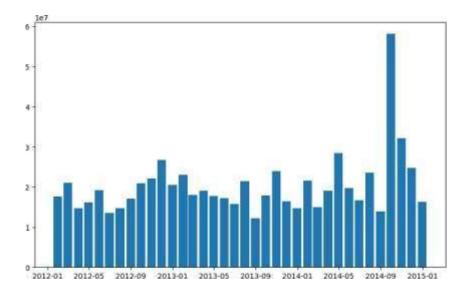
Resampling: Resampling is a methodology of economically using a data sample to improve the accuracy and quantify the uncertainty of a population parameter. Resampling for months or

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')
```

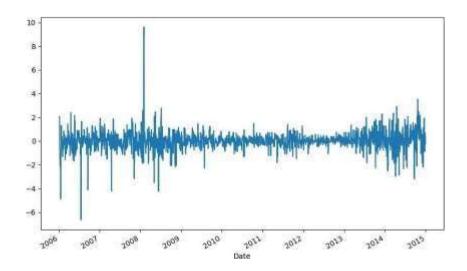
```
Ŧ
                               High
                    0pen
                                        Low
                                                  Close
                                                           Volume
                                                                    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>

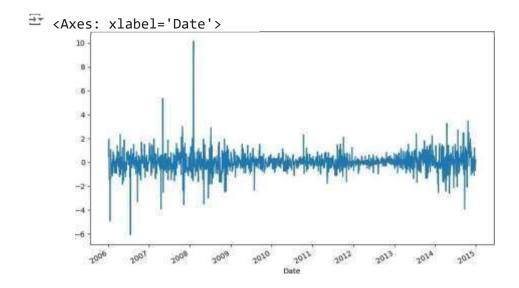


Differencing: Differencing is used to make the difference in values of a specified interval. By default, it's one, we can specify different values for plots. It is the most popular method to remove trends in the data.

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



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



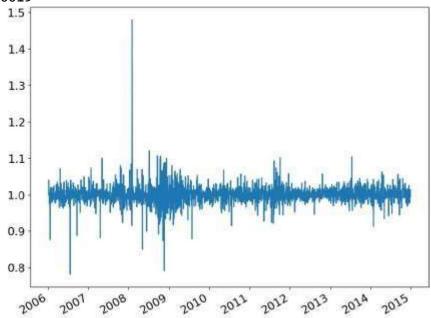
Plotting the Changes in Data

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)



.div() function helps to fill up the missing data values. Actually, div() means 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))

