Experiment 6

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a) Different types of plots using Numpy and Matplotlob

b)Basic operations using pandas like series, data frames, indexing, filtering, combining and merging data frames.

THEORY:

- Matplotlib is a plotting library for Python. It is used along with NumPy to provide an environment that is an effective open source alternative for MatLab. It can also be used with graphics toolkits like PyQt and wxPython
- Matplotlib is the most powerful visualization library. To use it, you need to import its sub package pyplot.
- To get something productive from that data, we need to do some statistical analysis on it, and Numpy is a great tool for that. The various statistical operations include mean, median, standard deviation, variance, etc.
- All these operations can either be done on the whole array, or can be done on just the rows or the columns which can be observed on below jupyter notebook.

Pandas Data Visualization

Frason Francis / SE-IT / 201903020

```
In [81]:
                                                                                                 M
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
In [82]:
                                                                                                 H
from numpy.random import randn, randint, uniform, sample
In [85]:
df = pd.DataFrame(randn(1000), index = pd.date_range('2019-06-07', periods = 1000), columns
ts = pd.Series(randn(1000), index = pd.date_range('2019-06-07', periods = 1000))
df.head()
Out[85]:
              value
            1.232051
2019-06-07
2019-06-08
           -0.261482
2019-06-09
            1.242395
2019-06-10
           0.895954
 2019-06-11
           0.736072
In [87]:
                                                                                                 H
df['value'] = df['value'].cumsum()
df.head()
Out[87]:
               value
2019-06-07
            1.232051
2019-06-08
            2.202620
2019-06-09
            4.415583
2019-06-10
            7.524501
 2019-06-11 11.369490
```

```
In [88]:

ts = ts.cumsum()
ts.head()
```

Out[88]:

2019-06-07 -0.173462 2019-06-08 -1.410669 2019-06-09 -1.760961 2019-06-10 -1.244528 2019-06-11 -0.363838 Freq: D, dtype: float64

In [89]: ▶

```
type(df), type(ts)
```

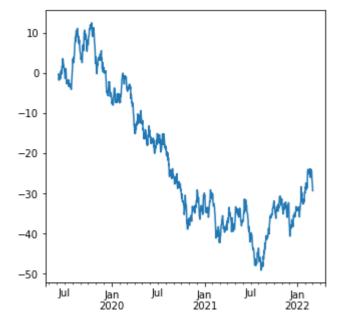
Out[89]:

(pandas.core.frame.DataFrame, pandas.core.series.Series)

In [92]:
ts.plot(figsize=(5,5))

Out[92]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2ac85a20>

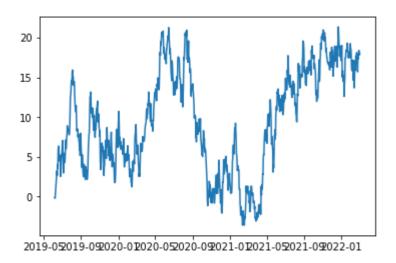


In [8]: ▶

plt.plot(ts)

Out[8]:

[<matplotlib.lines.Line2D at 0x1a1ac73668>]

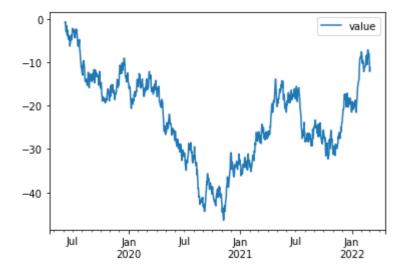


In [9]:

df.plot()

Out[9]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1aa34160>



In [10]:

```
iris = sns.load_dataset('iris')
iris.head()
```

Out[10]:

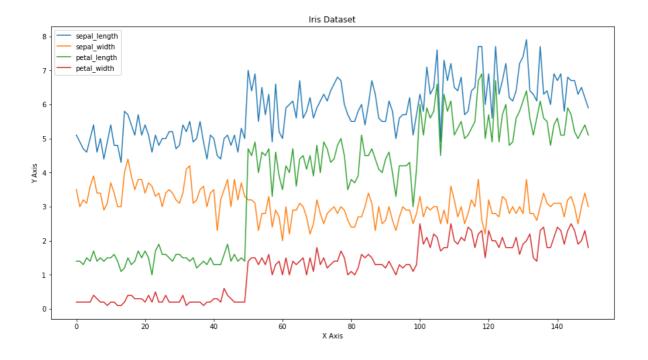
| | sepal_length | sepal_width | petal_length | petal_width | species |
|---|--------------|-------------|--------------|-------------|---------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 | setosa |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 | setosa |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 | setosa |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 | setosa |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 | setosa |

```
In [11]:
```

```
ax = iris.plot(figsize=(15,8), title='Iris Dataset')
ax.set_xlabel('X Axis')
ax.set_ylabel('Y Axis')
```

Out[11]:

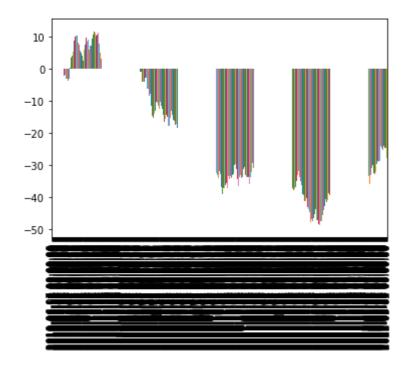
Text(0, 0.5, 'Y Axis')



```
In [94]:
ts.plot(kind = 'bar')
```

Out[94]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2ae7d080>



```
In [95]:

df = iris.drop(['species'], axis = 1)

In [97]:

df.iloc[0]
```

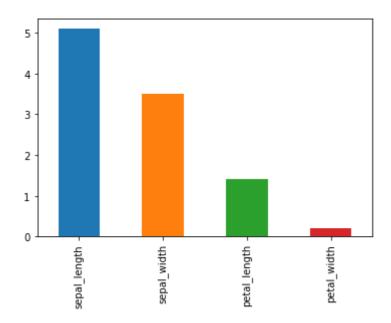
Out[97]:

sepal_length 5.1
sepal_width 3.5
petal_length 1.4
petal_width 0.2
Name: 0, dtype: float64

df.iloc[0].plot(kind='bar')

Out[15]:

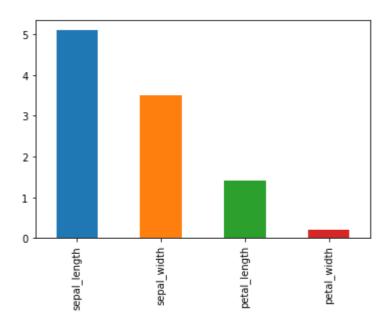
<matplotlib.axes._subplots.AxesSubplot at 0x1a1bbc4dd8>





Out[16]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1c99e940>



In [98]:

```
titanic = sns.load_dataset('titanic')
```

In [99]:

titanic.head()

Out[99]:

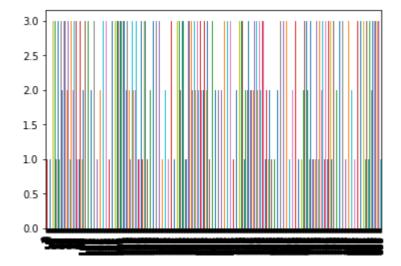
| | survived | pclass | sex | age | sibsp | parch | fare | embarked | class | who | adult_male |
|---|----------|--------|--------|------|-------|-------|---------|----------|-------|-------|------------|
| 0 | 0 | 3 | male | 22.0 | 1 | 0 | 7.2500 | S | Third | man | True |
| 1 | 1 | 1 | female | 38.0 | 1 | 0 | 71.2833 | С | First | woman | False |
| 2 | 1 | 3 | female | 26.0 | 0 | 0 | 7.9250 | S | Third | woman | False |
| 3 | 1 | 1 | female | 35.0 | 1 | 0 | 53.1000 | S | First | woman | False |
| 4 | 0 | 3 | male | 35.0 | 0 | 0 | 8.0500 | S | Third | man | True |
| 4 | | | | | | | | | | | • |

In [100]:

titanic['pclass'].plot(kind = 'bar')

Out[100]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a22dee6d8>



```
In [101]:

df = pd.DataFrame(randn(10, 4), columns=['a', 'b', 'c', 'd'])
df.head(10)
```

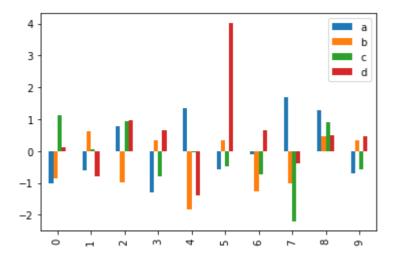
Out[101]:

| | а | b | С | d |
|---|-----------|-----------|-----------|-----------|
| 0 | 0.326203 | 2.331026 | -0.454617 | 0.107559 |
| 1 | 0.468923 | 1.185780 | 0.008878 | 0.723547 |
| 2 | 2.054247 | 1.828960 | 1.536323 | -1.792616 |
| 3 | 0.170623 | 0.640836 | 1.402193 | 0.045841 |
| 4 | 0.009997 | -0.727844 | 0.079510 | -1.533088 |
| 5 | -0.197923 | 0.135551 | 1.871942 | 1.361573 |
| 6 | 0.798528 | -0.079833 | 1.438415 | 0.397582 |
| 7 | 0.995109 | -1.384738 | -0.012644 | -1.937791 |
| 8 | 1.436894 | -0.254240 | -0.760523 | -0.523546 |
| 9 | -0.210206 | -0.766180 | -1.179217 | -1.280725 |

| In [21]: | H |
|---------------|---|
| df.plot.bar() | |

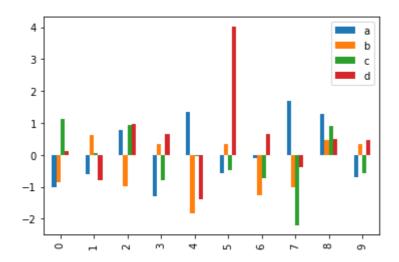
Out[21]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1c994860>



Out[22]:

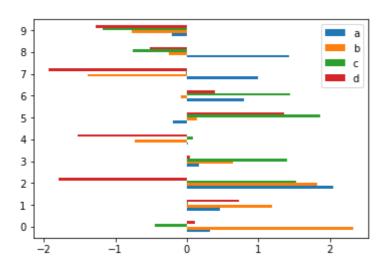
<matplotlib.axes._subplots.AxesSubplot at 0x1a1c9dc9e8>





Out[102]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2b6de208>

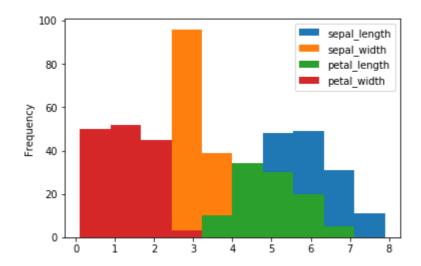


In [103]: ▶

iris.plot.hist()

Out[103]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2c1e9c18>

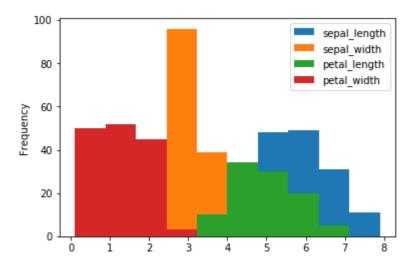


In [25]: ▶

iris.plot(kind = 'hist')

Out[25]:

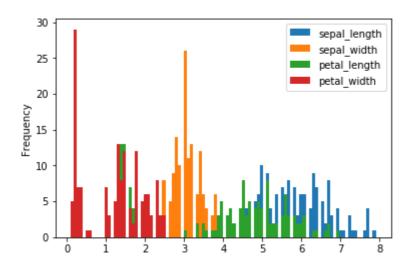
<matplotlib.axes._subplots.AxesSubplot at 0x1a1d22ec50>



```
iris.plot(kind = 'hist', stacked = False, bins = 100)
```

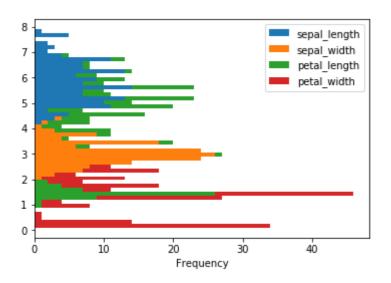
Out[110]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2d5a3b70>



Out[111]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2da82320>



H In [112]:

```
iris['sepal_width'].diff()
```

Out[112]:

```
0
       NaN
1
      -0.5
2
       0.2
3
      -0.1
4
       0.5
5
       0.3
6
      -0.5
7
       0.0
8
      -0.5
9
       0.2
10
       0.6
11
      -0.3
12
      -0.4
13
       0.0
14
       1.0
15
       0.4
      -0.5
16
17
      -0.4
18
       0.3
19
       0.0
20
      -0.4
21
       0.3
22
      -0.1
23
      -0.3
24
       0.1
25
      -0.4
26
       0.4
27
       0.1
28
      -0.1
29
      -0.2
       . . .
120
       1.0
121
      -0.4
122
       0.0
123
      -0.1
124
       0.6
125
      -0.1
126
      -0.4
       0.2
127
128
      -0.2
129
       0.2
130
      -0.2
131
       1.0
132
      -1.0
133
       0.0
134
      -0.2
135
       0.4
136
       0.4
137
      -0.3
138
      -0.1
139
       0.1
140
       0.0
141
       0.0
142
      -0.4
```

```
143  0.5

144  0.1

145  -0.3

146  -0.5

147  0.5

148  0.4

149  -0.4

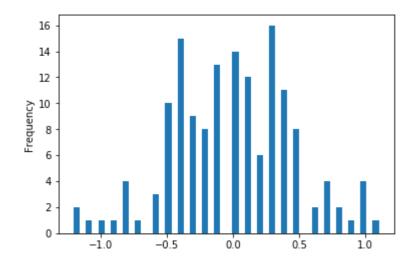
Name: sepal_width, Length: 150, dtype: float64
```

In [113]:

```
iris['sepal_width'].diff().plot(kind = 'hist', stacked = True, bins = 50)
```

Out[113]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2dcbc198>



In []: ▶

In [118]:

df = iris.drop(['species'], axis = 1)
df.diff().head()

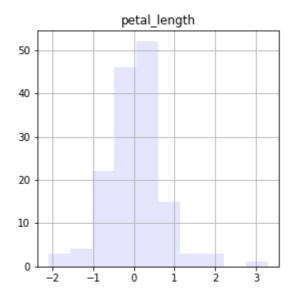
Out[118]:

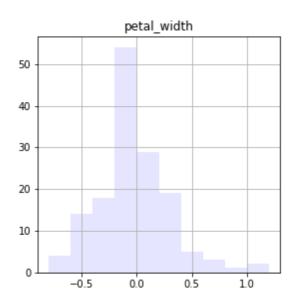
| | sepal_length | sepal_width | petal_length | petal_width |
|---|--------------|-------------|--------------|-------------|
| 0 | NaN | NaN | NaN | NaN |
| 1 | -0.2 | -0.5 | 0.0 | 0.0 |
| 2 | -0.2 | 0.2 | -0.1 | 0.0 |
| 3 | -0.1 | -0.1 | 0.2 | 0.0 |
| 4 | 0.4 | 0.5 | -0.1 | 0.0 |

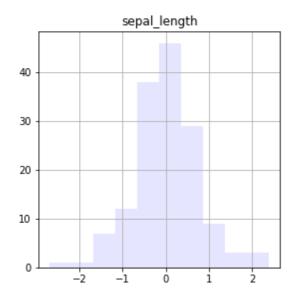
In [121]:

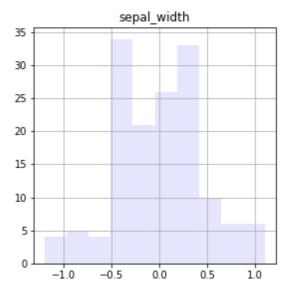
```
df.diff().hist(color = 'b', alpha = 0.1, figsize=(10,10))
```

Out[121]:









```
In [122]:

color = {'boxes': 'DarkGreen', 'whiskers': 'b'}

Out[122]:
{'boxes': 'DarkGreen', 'whiskers': 'b'}

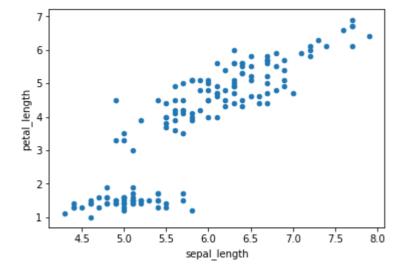
In []:

M

df.plot.scatter(x = 'sepal_length', y = 'petal_length')
```

Out[123]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2ebce518>

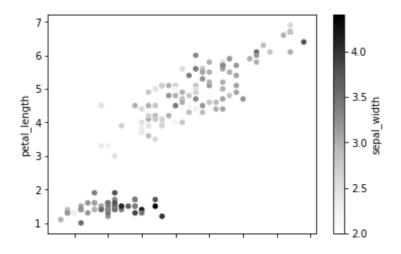


In [125]: ▶

```
df.plot.scatter(x = 'sepal_length', y = 'petal_length', c = 'sepal_width')
```

Out[125]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2ef4ec18>



In [126]: ▶

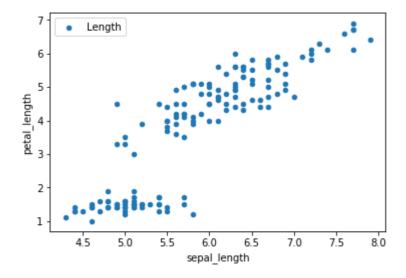
df.head()

Out[126]:

| | sepal_length | sepal_width | petal_length | petal_width |
|---|--------------|-------------|--------------|-------------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 |

```
In [131]:
```

```
df.plot.scatter(x = 'sepal_length', y = 'petal_length', label = 'Length');
#df.plot.scatter(x = 'sepal_width', y = 'petal_width', label = 'Width', ax = ax, color = 'r
#df.plot.scatter(x = 'sepal_width', y = 'petal_length', label = 'Width', ax = ax, color = '
```



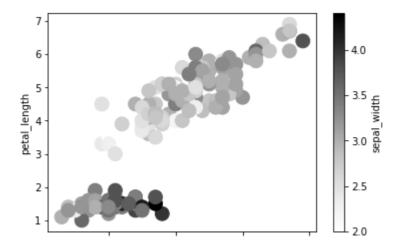
```
In [ ]: ▶
```

```
In [133]:

df.plot.scatter(x = 'sepal_length', y = 'petal_length', c = 'sepal_width', s = 190)
```

Out[133]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2f5fde80>



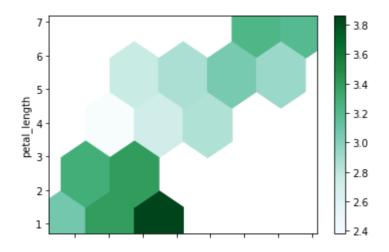
In []:

In [135]:

```
df.plot.hexbin(x = 'sepal_length', y = 'petal_length', gridsize = 5, C = 'sepal_width')
```

Out[135]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a2f883b00>



```
In [ ]: ▶
```

In [136]:

```
d = df.iloc[0]
d
```

Out[136]:

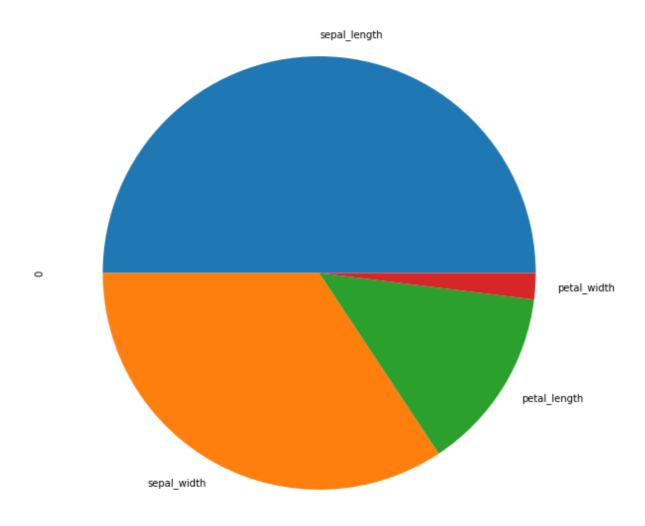
sepal_length 5.1
sepal_width 3.5
petal_length 1.4
petal_width 0.2
Name: 0, dtype: float64

In [39]: ▶

d.plot.pie(figsize = (10,10))

Out[39]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1e0fea58>



In [137]: ▶

d = df.head(3).T

In [138]: ▶

d

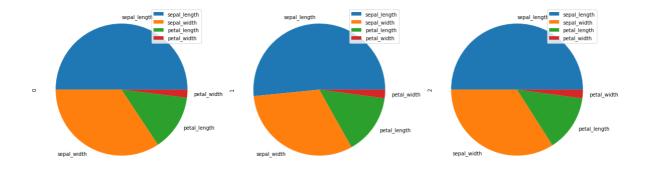
Out[138]:

| | 0 | 1 | 2 |
|--------------|-----|-----|-----|
| sepal_length | 5.1 | 4.9 | 4.7 |
| sepal_width | 3.5 | 3.0 | 3.2 |
| petal_length | 1.4 | 1.4 | 1.3 |
| petal_width | 0.2 | 0.2 | 0.2 |

In [140]:

```
d.plot.pie(subplots = True, figsize = (20, 20))
```

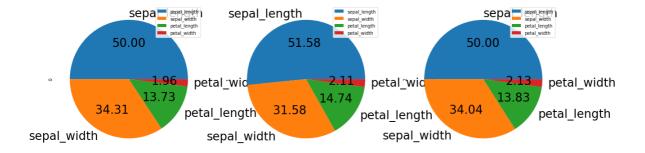
Out[140]:



In [142]: ▶

```
d.plot.pie(subplots = True, figsize = (20, 20), fontsize = 26, autopct = '%.2f')
```

Out[142]:



In [44]: ▶

```
[0.1]*4
```

Out[44]:

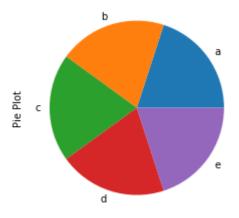
[0.1, 0.1, 0.1, 0.1]

In [144]: ▶

```
series = pd.Series([0.2]*5, index = ['a','b','c', 'd','e'], name = 'Pie Plot')
series.plot.pie()
```

Out[144]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a306f4ba8>



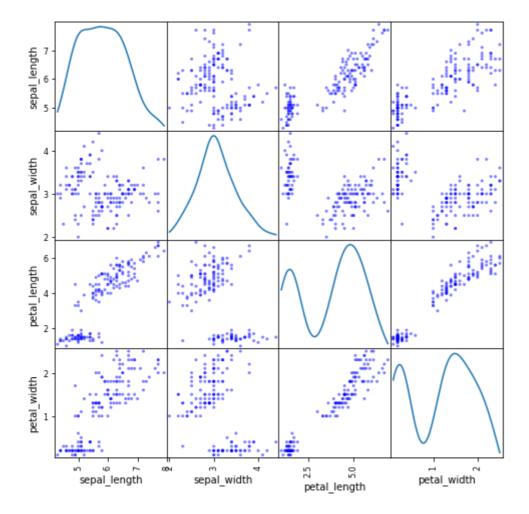
In []:

```
In [46]: ▶
```

from pandas.plotting import scatter_matrix

In [47]: ▶

scatter_matrix(df, figsize= (8,8), diagonal='kde', color = 'b')
plt.show()

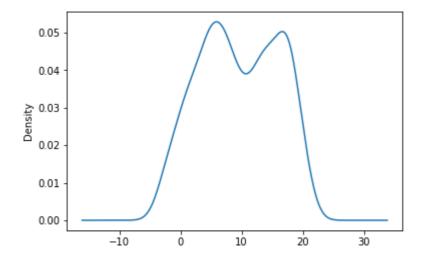


In [48]:

ts.plot.kde()

Out[48]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1f41df98>



In [49]: ▶

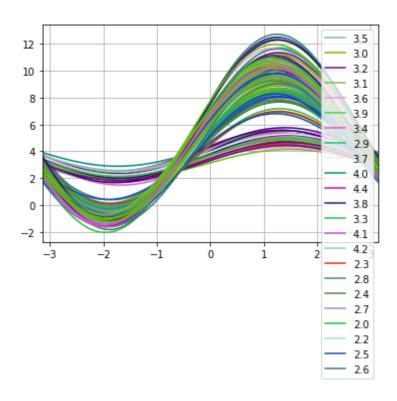
from pandas.plotting import andrews_curves

In [50]: ▶

andrews_curves(df, 'sepal_width')

Out[50]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1eda8dd8>



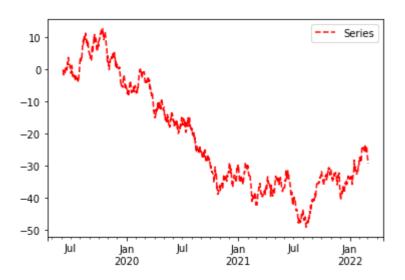
In []: ▶

In [145]:

ts.plot(style = 'r--', label = 'Series', legend = True)

Out[145]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a30949f28>

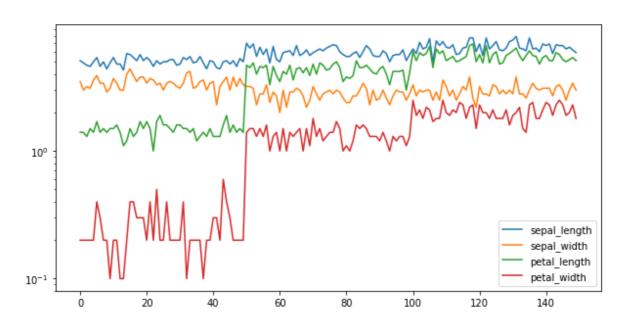


In [146]:

```
df.plot(legend = True, figsize = (10, 5), logy = True)
```

Out[146]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a30a13ac8>



In [53]: ▶

df.head(0)

Out[53]:

sepal_length sepal_width petal_length petal_width

In [54]: ▶

```
x = df.drop(['sepal_width', 'petal_width'], axis = 1)
x.head()
```

Out[54]:

| | sepal_length | petal_length |
|---|--------------|--------------|
| 0 | 5.1 | 1.4 |
| 1 | 4.9 | 1.4 |
| 2 | 4.7 | 1.3 |
| 3 | 4.6 | 1.5 |
| 4 | 5.0 | 1.4 |

In [55]:

```
y = df.drop(['sepal_length', 'petal_length'], axis = 1)
y.head()
```

Out[55]:

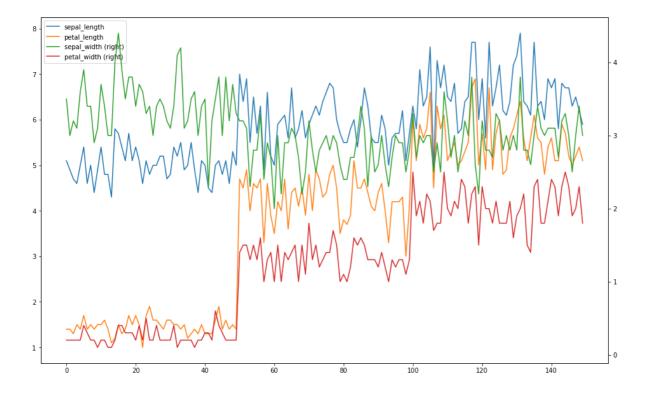
| | sepal_width | petal_width |
|---|-------------|-------------|
| 0 | 3.5 | 0.2 |
| 1 | 3.0 | 0.2 |
| 2 | 3.2 | 0.2 |
| 3 | 3.1 | 0.2 |
| 4 | 3.6 | 0.2 |

In [56]: ▶

```
ax = x.plot()
y.plot(figsize = (16,10), secondary_y=True, ax = ax)
```

Out[56]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1e6d7e80>

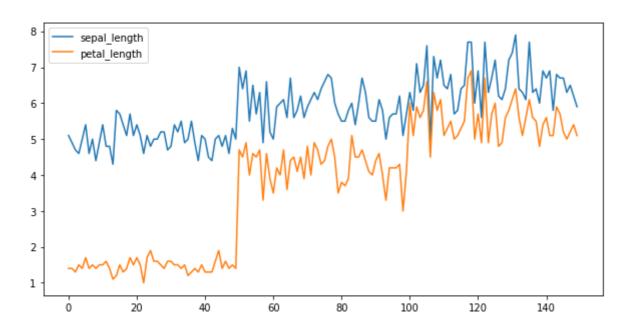


In [57]: ▶

x.plot(figsize=(10,5), x_compat = True)

Out[57]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1e6c3a58>

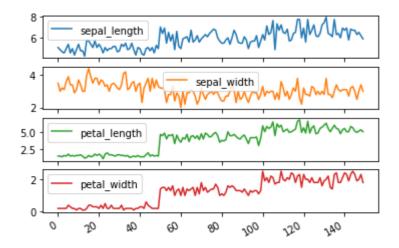


In []:

In [58]:

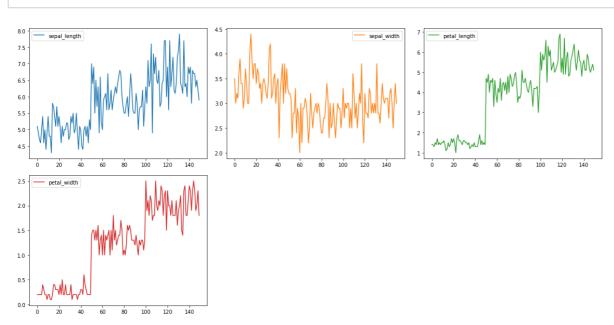
```
df.plot(subplots = True)
```

Out[58]:



In [59]: ▶

```
df.plot(subplots = True, sharex = False, layout = (2,3), figsize = (16,8))
plt.tight_layout()
```



In [78]: ▶

from descartes import PolygonPatch

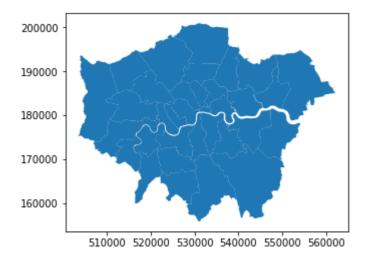
In []:

In [79]:

map_df = gpd.read_file("London_Borough_Excluding_MHW.shp")
check data type so we can see that this is not a normal dataframe, but a GEOdataframe
map_df.head()
map_df.plot()

Out[79]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a1e9f9630>



In [71]:

import geopandas as gpd

In [80]: ▶

map_df

Out[80]:

| | NAME | GSS_CODE | HECTARES | NONLD_AREA | ONS_INNER | SUB_2009 | SUB_2006 |
|---|-------------------------|-----------|-----------|------------|-----------|----------|----------|
| 0 | Kingston upon Thames | E09000021 | 3726.117 | 0.000 | F | None | None |
| 1 | Croydon | E09000008 | 8649.441 | 0.000 | F | None | None |
| 2 | Bromley | E09000006 | 15013.487 | 0.000 | F | None | None |
| 3 | Hounslow | E09000018 | 5658.541 | 60.755 | F | None | None |
| 4 | Ealing | E09000009 | 5554.428 | 0.000 | F | None | None |
| 5 | Havering | E09000016 | 11445.735 | 210.763 | F | None | None |
| 6 | Hillingdon | E09000017 | 11570.063 | 0.000 | F | None | None |
| 7 | Harrow | E09000015 | 5046.330 | 0.000 | F | None | None |
| 8 | Brent | E09000005 | 4323.270 | 0.000 | F | None | None |
| 9 | Barnet | E09000003 | 8674.837 | 0.000 | F | None | None |

| | NAME | GSS_CODE | HECTARES | NONLD_AREA | ONS_INNER | SUB_2009 | SUB_2006 |
|----|-------------------------|-----------|----------|------------|-----------|----------|----------|
| 10 | Lambeth | E09000022 | 2724.940 | 43.927 | Т | None | None |
| 11 | Southwark | E09000028 | 2991.340 | 105.139 | Т | None | None |
| 12 | Lewisham | E09000023 | 3531.706 | 16.795 | Т | None | None |
| 13 | Greenwich | E09000011 | 5044.190 | 310.785 | F | None | None |
| 14 | Bexley | E09000004 | 6428.649 | 370.619 | F | None | None |
| 15 | Enfield | E09000010 | 8220.025 | 0.000 | F | None | None |
| 16 | Waltham Forest | E09000031 | 3880.793 | 0.000 | F | None | None |
| 17 | Redbridge | E09000026 | 5644.225 | 2.300 | F | None | None |
| 18 | Sutton | E09000029 | 4384.698 | 0.000 | F | None | None |
| 19 | Richmond upon Thames | E09000027 | 5876.111 | 135.443 | F | None | None |
| 20 | Merton | E09000024 | 3762.466 | 0.000 | F | None | None |
| 21 | Wandsworth | E09000032 | 3522.022 | 95.600 | Т | None | None |

| | NAME | GSS_CODE | HECTARES | NONLD_AREA | ONS_INNER | SUB_2009 | SUB_2006 |
|----|---------------------------|-----------|----------|------------|-----------|----------|----------|
| 22 | Hammersmith and Fulham | E09000013 | 1715.409 | 75.648 | Т | None | None |
| 23 | Kensington and Chelsea | E09000020 | 1238.379 | 25.994 | Т | None | None |
| 24 | Westminster | E09000033 | 2203.005 | 54.308 | Т | None | None |
| 25 | Camden | E09000007 | 2178.932 | 0.000 | Т | None | None |
| 26 | Tower Hamlets | E09000030 | 2157.501 | 179.707 | Т | None | None |
| 27 | Islington | E09000019 | 1485.664 | 0.000 | Т | None | None |
| 28 | Hackney | E09000012 | 1904.902 | 0.000 | Т | None | None |
| 29 | Haringey | E09000014 | 2959.837 | 0.000 | Т | None | None |
| 30 | Newham | E09000025 | 3857.806 | 237.637 | Т | None | None |
| 31 | Barking and Dagenham | E09000002 | 3779.934 | 169.150 | F | None | None |
| 32 | City of London | E09000001 | 314.942 | 24.546 | Т | None | None |

| • |) | |
|---------|----------|---|
| In []: | | H |
| | | |
| In []: | | H |
| | | |

Numpy Operations on Array

By: Frason Francis - SE-IT:201903020

```
In [5]:
import numpy as np
type(np.array(["frason", 2, 3]))
Out[5]:
numpy.ndarray
                                                                                             H
In [2]:
#Upcasting:
In [3]:
np.array([1, 2, 3.0])
Out[3]:
array([1., 2., 3.])
In [4]:
                                                                                             H
#two dimensions
In [4]:
 np.array([[1, 2], [3, 4]])
Out[4]:
array([[1, 2],
       [3, 4]])
In [6]:
                                                                                             H
#Minimum dimensions 2:
In [6]:
np.array([1, 2, 3], ndmin=5)
Out[6]:
array([[[[[1, 2, 3]]]])
```

```
H
In [8]:
#dtype
In [7]:
                                                                                            H
 np.array([1, 2, 3], dtype=complex)
Out[7]:
array([1.+0.j, 2.+0.j, 3.+0.j])
In [12]:
                                                                                            H
#Data-type consisting of more than one element:
In [15]:
                                                                                            M
x = np.array([(1,2),(3,4)],dtype=[('a','<i2'),('b','<i8')])
Out[15]:
array([(1, 2), (3, 4)], dtype=[('a', '<i2'), ('b', '<i8')])
In [39]:
                                                                                            M
type(x[1][0])
Out[39]:
numpy.int16
                                                                                            H
In [13]:
#Creating an array from sub-classes:
In [17]:
np.mat(np.array([[1, 2],[4,7]]))
Out[17]:
matrix([[1, 2],
        [4, 7]])
In [21]:
                                                                                            H
np.mat('1 2; 3 4')
Out[21]:
matrix([[1, 2],
        [3, 4]])
```

numpy.asarray

```
H
In [16]:
#Convert the input to an array.
In [18]:
                                                                                             H
#Convert a list into an array:
In [27]:
a = [1, 2]
type(a)
Out[27]:
list
In [25]:
                                                                                             H
type(np.asarray(a))
Out[25]:
numpy.ndarray
In [31]:
                                                                                             H
a = np.array([1, 2]) #Existing arrays are not copied
type(a)
Out[31]:
numpy.ndarray
In [42]:
np.asarray((1,2))
Out[42]:
array([1, 2])
In [25]:
#If dtype is set, array is copied only if dtype does not match:
```

```
In [43]:
                                                                                             H
a = np.array([1, 2], dtype=np.float32)
Out[43]:
array([1., 2.], dtype=float32)
In [33]:
                                                                                             H
np.asarray([1,2]) is a
Out[33]:
False
                                                                                             H
In [36]:
np.asarray([1,2])
Out[36]:
array([1, 2])
In [34]:
                                                                                             H
np.asarray(a, dtype=np.float64) is a
Out[34]:
False
                                                                                             H
In [29]:
# ndarray subclasses are not passed through
In [44]:
issubclass(np.matrix, np.ndarray)
Out[44]:
True
In [48]:
                                                                                             H
a = np.matrix([[1, 2]])
np.asanyarray(a)
Out[48]:
matrix([[1, 2]])
```

```
H
In [49]:
np.asarray(a) is a
Out[49]:
False
In [50]:
                                                                                          H
np.asanyarray(a) is a
Out[50]:
True
                                                                                          H
In [51]:
type(np.asarray(a))
Out[51]:
numpy.ndarray
numpy.asanyarray
In [50]:
                                                                                          H
a = [1, 2]
np.asanyarray(a)
Out[50]:
array([1, 2])
                                                                                          H
In [86]:
a = np.matrix([1, 2])
а
Out[86]:
matrix([[1, 2]])
In [52]:
                                                                                          M
np.asanyarray([1,2]) is a
Out[52]:
True
#numpy.copy
```

```
In [52]:
np.array(a, copy=True)
Out[52]:
array([[1, 2]])
In [40]:
\#Create an array x, with a reference y and a copy z:
In [56]:
x = np.array([1, 2, 3])
Х
У
Out[56]:
array([1, 2, 3])
                                                                                           H
In [62]:
y = x
In [63]:
z = np.copy(x)
In [64]:
y[0] = 100
Out[64]:
array([100,
            2,
                   3])
In [78]:
Χ
Out[78]:
array([100, 2,
                   3])
```

```
H
In [89]:
id(x[0])
Out[89]:
4767545360
In [91]:
                                                                                         H
id(y[0])
Out[91]:
4767545408
In [ ]:
                                                                                         M
In [92]:
id(x)
Out[92]:
4768156128
In [93]:
                                                                                         H
id(y)
Out[93]:
4768156128
numpy.fromfunction
In [58]:
                                                                                         H
#Construct an array by executing a function over each coordinate.
In [94]:
np.fromfunction(lambda i, j: i == j, (3, 3), dtype=int)
Out[94]:
array([[ True, False, False],
       [False, True, False],
       [False, False, True]])
```

```
In [119]:
                                                                                            H
np.fromfunction(lambda i, j: i * j, (3, 3), dtype=int)
Out[119]:
array([[0, 0, 0],
       [0, 1, 2],
       [0, 2, 4]])
In [61]:
                                                                                            H
#Create a new 1-dimensional array from an iterable object.
In [97]:
                                                                                            H
iterable = (x*x for x in range(5))
iterable
Out[97]:
<generator object <genexpr> at 0x11c326b10>
In [98]:
                                                                                            M
np.fromiter(iterable, float)
Out[98]:
array([ 0., 1., 4., 9., 16.])
In [64]:
                                                                                            M
#A new 1-D array initialized from text data in a string
In [101]:
a = np.fromstring('234 234',sep=' ')
а
Out[101]:
array([234., 234.])
In [102]:
                                                                                            H
np.fromstring('1, 2', dtype=int, sep=',')
Out[102]:
array([1, 2])
In [79]:
                                                                                            H
# How to create create a record array from a (flat) list of arrays
```

```
In [103]:
                                                                                           H
x1=np.array([1,2,3,4])
In [116]:
                                                                                           M
x2=np.array(['a','dd','xyz','12'])
In [117]:
x3=np.array([1.1,2,3,4])
x4=np.array([1.1,2,3,4])
type(x4)
Out[117]:
numpy.ndarray
In [118]:
                                                                                           M
r = np.core.records.fromarrays([x1,x2,x3,x4],names='a,b,c,d')
r
Out[118]:
rec.array([(1, 'a', 1.1, 1.1), (2, 'dd', 2., 2.), (3, 'xyz', 3., 3.),
           (4, '12', 4., 4.)],
          dtype=[('a', '<i8'), ('b', '<U3'), ('c', '<f8'), ('d', '<f8')])
In [112]:
                                                                                           H
print(r[1]["a"])
2
x1[1]=34
In [130]:
x1[1]
Out[130]:
2
```

data types

```
In [120]:
my_list = [1,2,3]
import numpy as np
arr = np.array(my_list)
print("Type/Class of this object:",type(arr))
print("Here is the vector\n----\n",arr)
Type/Class of this object: <class 'numpy.ndarray'>
Here is the vector
 [1 2 3]
In [121]:
                                                                                        H
my_mat = [[1,2,3],[4,5,6],[7,8,9]]
mat = np.array(my_mat)
print("Type/Class of this object:",type(mat))
print("Here is the matrix\n-----\n",mat,"\n-----")
print("Dimension of this matrix: ",mat.ndim,sep='') #ndim gives the dimensison, 2 for a mat
print("Size of this matrix: ", mat.size,sep='') #size gives the total number of elements
print("Shape of this matrix: ", mat.shape, sep='') #shape gives the number of elements along
print("Data type of this matrix: ", mat.dtype, sep='') #dtype gives the data type contained
Type/Class of this object: <class 'numpy.ndarray'>
Here is the matrix
 [[1 2 3]
 [4 5 6]
 [7 8 9]]
-----
Dimension of this matrix: 2
Size of this matrix: 9
Shape of this matrix: (3, 3)
Data type of this matrix: int64
In [122]:
                                                                                        H
my_mat = [[1.1,2,3],[4,5.2,6],[7,8.3,9]]
mat = np.array(my mat)
print("Data type of the modified matrix: ", mat.dtype, sep='') #dtype gives the data type co
print("\n\nEven tuples can be converted to ndarrays...")
Data type of the modified matrix: float64
```

Even tuples can be converted to ndarrays...

```
In [111]:
b = np.array([(1.5,2,3), (4,5,6)])
print("We write b = np.array([(1.5,2,3), (4,5,6)])")
print("Matrix made from tuples, not lists\n----")
print(b)
We write b = np.array([(1.5,2,3), (4,5,6)])
Matrix made from tuples, not lists
[[1.5 2. 3.]
 [4. 5. 6.]]
arange and linspace
                                                                                     M
In [127]:
print("A series of numbers:", type(np.arange(5,16)))
np.arange(5,16,2.3)# A series of numbers from low to high
A series of numbers: <class 'numpy.ndarray'>
Out[127]:
array([ 5. , 7.3, 9.6, 11.9, 14.2])
In [128]:
                                                                                     H
list(range(5,16,2))
Out[128]:
[5, 7, 9, 11, 13, 15]
In [129]:
                                                                                     H
list(range(50,-1,5))
Out[129]:
[]
In [135]:
                                                                                     H
print("Numbers spaced apart by 2:",np.arange(50,-1,5)) # Numbers spaced apart by 2
Numbers spaced apart by 2: []
In [131]:
print("Numbers spaced apart by float:",np.arange(0,11,2.5)) # Numbers spaced apart by 2.5
```

7.5 10.]

Numbers spaced apart by float: [0. 2.5 5.

```
In [132]:
print("Every 5th number from 50 in reverse order\n",np.arange(5.0,-1,-5))
Every 5th number from 50 in reverse order
[5. 0.]
In [134]:
print("21 linearly spaced numbers between 1 and 5\n-----
print((np.linspace(1,5,50)))
21 linearly spaced numbers between 1 and 5
-----
[1.
           1.08163265 1.16326531 1.24489796 1.32653061 1.40816327
1.48979592 1.57142857 1.65306122 1.73469388 1.81632653 1.89795918
1.97959184 2.06122449 2.14285714 2.2244898 2.30612245 2.3877551
2.46938776 2.55102041 2.63265306 2.71428571 2.79591837 2.87755102
2.95918367 3.04081633 3.12244898 3.20408163 3.28571429 3.36734694
3.44897959 3.53061224 3.6122449 3.69387755 3.7755102 3.85714286
3.93877551 4.02040816 4.10204082 4.18367347 4.26530612 4.34693878
4.42857143 4.51020408 4.59183673 4.67346939 4.75510204 4.83673469
4.91836735 5.
                    ]
Matrix creation
```

```
In [127]:
                                                                                H
print("Vector of ones\n----")
print(np.ones(5))
Vector of ones
[1. 1. 1. 1. 1.]
In [140]:
                                                                                M
print("Matrix of ones\n----")
print(np.ones((5,2,8))) # Note matrix dimension specified by Tuples
Matrix of ones
-----
[[[1. 1. 1. 1. 1. 1. 1. 1.]
  [1. 1. 1. 1. 1. 1. 1. ]
 [[1. 1. 1. 1. 1. 1. 1. 1.]
 [1. 1. 1. 1. 1. 1. 1. 1.]]
 [[1. 1. 1. 1. 1. 1. 1. 1.]
 [1. 1. 1. 1. 1. 1. 1. 1.]]
 [[1. 1. 1. 1. 1. 1. 1. 1.]
 [1. 1. 1. 1. 1. 1. 1. ]
 [[1. 1. 1. 1. 1. 1. 1. 1.]
 [1. 1. 1. 1. 1. 1. 1. 1.]]]
In [147]:
                                                                                M
print("Matrix of 5's\n----")
print(5+np.ones((3,5)))
Matrix of 5's
------
[[6. 6. 6. 6. 6.]
[6. 6. 6. 6. 6.]
[6. 6. 6. 6. 6.]]
                                                                                H
In [148]:
print("Empty matrix\n----\n", np.empty((3,5)))
Empty matrix
-----
 [[6. 6. 6. 6. 6.]
 [6. 6. 6. 6. 6.]
 [6. 6. 6. 6. 6.]]
```

```
In [149]:
                                                                                             H
mat1 = np.eye(4)
print("Identity matrix of dimension", mat1.shape)
print(mat1)
Identity matrix of dimension (4, 4)
[[1. 0. 0. 0.]
 [0. 1. 0. 0.]
 [0. 0. 1. 0.]
 [0. 0. 0. 1.]]
In [132]:
                                                                                             M
np.arange(3)
Out[132]:
array([0, 1, 2])
In [80]:
                                                                                             M
np.arange(3.0)
Out[80]:
array([0., 1., 2.])
In [81]:
                                                                                             M
 np.arange(3,7)
Out[81]:
array([3, 4, 5, 6])
In [82]:
                                                                                             H
np.arange(3,7,2)
Out[82]:
array([3, 5])
In [150]:
                                                                                             H
np.linspace(2.0, 3.0, num=5)
Out[150]:
array([2. , 2.25, 2.5 , 2.75, 3. ])
```

```
In [151]:
                                                                                        H
np.linspace(2.0, 3.0, num=5, endpoint=False)
Out[151]:
array([2., 2.2, 2.4, 2.6, 2.8])
In [153]:
                                                                                        H
np.linspace(2.0, 3.0, num=9, retstep=True)
Out[153]:
(array([2. , 2.125, 2.25 , 2.375, 2.5 , 2.625, 2.75 , 2.875, 3. ]), 0.1
25)
                                                                                        H
In [147]:
#Return numbers spaced evenly on a log scale.
np.linspace(2.0, 3.0, num=4)
Out[147]:
           , 2.33333333, 2.66666667, 3.
array([2.
                                                    1)
In [154]:
                                                                                        M
np.logspace(2.0, 3.0, num=4, base = 10)
Out[154]:
array([ 100.
            , 215.443469 , 464.15888336, 1000.
                                                                ])
In [113]:
                                                                                        H
np.logspace(2.0, 3.0, num=4, endpoint=False)
Out[113]:
array([ 100. , 177.827941 , 316.22776602, 562.34132519])
In [155]:
                                                                                        H
np.logspace(2.0, 3.0, num=4, base=2.0)
Out[155]:
            , 5.0396842 , 6.34960421, 8.
                                                    ])
array([4.
In [115]:
#Extract a diagonal or construct a diagonal array.
```

```
H
In [179]:
x = np.arange(16).reshape((-1,4))
In [180]:
Х
Out[180]:
array([[ 0, 1, 2, 3],
       [4, 5, 6, 7],
       [8, 9, 10, 11],
       [12, 13, 14, 15]])
In [172]:
                                                                                         H
np.diag(x)
Out[172]:
array([ 0, 5, 10, 15])
                                                                                         H
In [182]:
np.diag(x, k=2)
Out[182]:
array([2, 7])
In [183]:
                                                                                         M
np.diag(x, k=-1)
Out[183]:
array([ 4, 9, 14])
In [184]:
np.diag(np.diag(x))
Out[184]:
array([[ 0, 0, 0, 0],
       [0, 5, 0, 0],
       [ 0, 0, 10, 0],
       [ 0, 0, 0, 15]])
In [122]:
                                                                                         H
#Create a two-dimensional array with the flattened input as a diagonal.
```

```
In [180]:
                                                                                             H
np.diagflat([[1,2], [3,4]])
Out[180]:
array([[1, 0, 0, 0],
       [0, 2, 0, 0],
       [0, 0, 3, 0],
       [0, 0, 0, 4]])
                                                                                             H
In [124]:
np.diagflat([1,2], 1)
Out[124]:
array([[0, 1, 0],
       [0, 0, 2],
       [0, 0, 0]])
In [125]:
                                                                                             M
#An array with ones at and below the given diagonal and zeros elsewhere.
In [187]:
                                                                                             H
np.tri(3, 5, 1, dtype=int)
Out[187]:
array([[1, 1, 0, 0, 0],
       [1, 1, 1, 0, 0],
       [1, 1, 1, 1, 0]])
In [191]:
                                                                                             H
np.tri(3, 5, k=-1)
Out[191]:
array([[0., 0., 0., 0., 0.],
       [1., 0., 0., 0., 0.],
       [1., 1., 0., 0., 0.]])
In [128]:
#return a Lower triangle of an array.
```

```
In [205]:
                                                                                      H
np.tril([[1,2,3],[4,5,6],[7,8,9]], 0)
Out[205]:
array([[1, 0, 0],
      [4, 5, 0],
      [7, 8, 9]])
In [130]:
                                                                                      H
#return Upper triangle of an array.
In [206]:
                                                                                      H
np.triu([[1,2,3],[4,5,6],[7,8,9],[10,11,12]], 1)
Out[206]:
array([[0, 2, 3],
      [0, 0, 6],
      [0, 0, 0],
      [0, 0, 0]])
Random number generation
In [197]:
                                                                                      H
print("Random number generation (from Uniform distribution)")
print(np.random.rand(2,3)) # 2 by 3 matrix with random numbers ranging from 0 to 1, Note no
Random number generation (from Uniform distribution)
[[0.79234365 0.26636652 0.84876143]
 [0.78766627 0.06627396 0.8829639 ]]
In [209]:
                                                                                      H
print("Numbers from Normal distribution with zero mean and standard deviation 1 i.e. standa
print(np.random.randn(4,3))
Numbers from Normal distribution with zero mean and standard deviation 1 i.
e. standard normal
[[ 0.08550888  0.73651756  0.08329424]
 [ 0.75179279  0.42441618  0.47752353]
 [ 0.48190152 -0.24940656 -0.44992115]]
```

```
In [202]:
print("Random integer vector:",np.random.randint(1,10)) #randint (low, high, # of samples t
print ("\nRandom integer matrix")
Random integer vector: 4
Random integer matrix
In [208]:
print(np.random.randint(1,100,(4,4))) #randint (low, high, # of samples to be drawn in a tu
print("\n20 samples drawn from a dice throw:",np.random.randint(1,7,20)) # 20 samples drawn
[[28 90 28 25]
 [14 51 32 98]
 [10 98 11 71]
 [12 42 54 77]]
20 samples drawn from a dice throw: [1 6 2 1 1 1 5 2 5 1 1 5 5 1 6 6 5 4 1
1]
Reshaping
In [211]:
                                                                                          M
from numpy.random import randint as ri
a = ri(1,100,30)
b = a.reshape(2,3,5)
c = a.reshape(6, -19878)
Out[211]:
array([[51, 63, 51, 54, 96],
       [11, 93, 2, 51, 34],
       [62, 35, 42, 95, 58],
       [88, 44, 36, 93, 83],
       [86, 56, 30, 2, 15],
       [68, 40, 77, 48, 57]])
In [209]:
print ("Shape of a:", a.shape)
print ("Shape of b:", b.shape)
print ("Shape of c:", c.shape)
Shape of a: (30,)
Shape of b: (2, 3, 5)
Shape of c: (6, 5)
```

```
In [219]:
print("\na looks like\n",'-'*20,"\n",a,"\n",'-'*20)
print("\nb looks like\n",'-'*20,"\n",b,"\n",'-'*20)
print("\nc looks like\n",'-'*20,"\n",c,"\n",'-'*20)
a looks like
[72 69 61 67 42 2 92 42 38 56 22 71 21 81 81 47 2 1 6 94 52 14 87 71
38 57 99 87 62 88]
 _____
b looks like
 -----
[[[72 69 61 67 42]
 [ 2 92 42 38 56]
 [22 71 21 81 81]]
 [[47 2 1 6 94]
 [52 14 87 71 38]
 [57 99 87 62 88]]]
 ______
c looks like
 -----
[[72 69 61 67 42]
[ 2 92 42 38 56]
[22 71 21 81 81]
[47 2 1 6 94]
 [52 14 87 71 38]
[57 99 87 62 88]]
In [212]:
                                                                                H
A = ri(1,100,10) \# Vector of random interegrs
print("\nVector of random integers\n",'-'*50,"\n",A)
print("\nHere is the sorted vector\n",'-'*50,"\n",np.sort(A))
Vector of random integers
 -----
[69 99 45 72 41 16 50 80 80 22]
Here is the sorted vector
 [16 22 41 45 50 69 72 80 80 99]
```

```
In [221]:
                                                                                           H
M = ri(1,100,25).reshape(5,5) # Matrix of random interegrs
#print("\nHere is the sorted matrix along each row\n",'-'*50,"\n",np.sort(M, kind='mergesor
print("\nHere is the sorted matrix along each column\n",'-'*50,"\n",np.sort(M, axis=1, kind
Here is the sorted matrix along each column
 [[18 25 38 73 94]
 [18 42 75 79 91]
 [21 44 62 78 85]
 [15 17 44 89 91]
 [ 3 24 38 62 86]]
Out[221]:
array([[18, 94, 25, 38, 73],
       [75, 42, 79, 18, 91],
       [21, 44, 85, 78, 62],
       [91, 15, 17, 44, 89],
       [62, 38, 3, 24, 86]])
In [214]:
print("Max of a:", M.max())
print("Max of b:", b.max())
Max of a: 99
Max of b: 96
Out[214]:
array([[[51, 63, 51, 54, 96],
        [11, 93, 2, 51, 34],
        [62, 35, 42, 95, 58]],
       [[88, 44, 36, 93, 83],
        [86, 56, 30, 2, 15],
        [68, 40, 77, 48, 57]]])
In [217]:
Μ
Out[217]:
array([[46, 88, 23, 66, 57],
       [50, 90, 30, 46, 36],
```

[31, 19, 42, 23, 31], [4, 44, 86, 10, 81], [73, 35, 30, 99, 43]])

```
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                                         Numpy-Operations - Jupyter Notebook
  In [218]:
  print("Max of a location:", M.argmax(axis= 1 ))
  print("Max of b location:", b.argmax())
  print("Max of c location:", b.argmax())
 Max of a location: [1 1 2 2 3]
 Max of b location: 4
 Max of c location: 4
  Indexing and slicing
                                                                                              H
  In [229]:
  arr = np.arange(0,11)
  print("Array:",arr)
```

```
Array: [ 0 1 2 3 4 5 6 7 8 9 10]
                                                                                       H
In [161]:
print("Element at 7th index is:", arr[7])
Element at 7th index is: 7
In [230]:
print("Elements from 3rd to 5th index are:", arr[3:6:2])
Elements from 3rd to 5th index are: [3 5]
                                                                                       H
In [231]:
print("Elements up to 4th index are:", arr[:4])
Elements up to 4th index are: [0 1 2 3]
Out[231]:
array([ 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10])
In [233]:
print("Elements from last backwards are:", arr[-1:7:-1])
```

Elements from last backwards are: [10 9 8]

```
In [164]:
                                                                                     H
print("3 Elements from last backwards are:", arr[-1:-6:2])
3 Elements from last backwards are: []
In [165]:
                                                                                     H
arr = np.arange(0,21,2)
print("New array:",arr)
New array: [ 0 2 4 6 8 10 12 14 16 18 20]
In [64]:
print("Elements at 2nd, 4th, and 9th index are:", arr[[2,4,9]]) # Pass a list as a index to
Elements at 2nd, 4th, and 9th index are: [ 4 8 18]
In [237]:
                                                                                     M
import numpy as np
mat = np.array(ri(10,100,15)).reshape(3,5)
print("Matrix of random 2-digit numbers\n----\n",mat)
mat[1:4,3:5]
mat[0:3,[1,3]]
Matrix of random 2-digit numbers
-----
 [[35 75 44 59 85]
 [71 58 75 72 69]
 [76 97 49 63 53]]
Out[237]:
array([[75, 59],
      [58, 72],
      [97, 63]])
In [238]:
mat[0:3,[1,3]]
mat
Out[238]:
array([[35, 75, 44, 59, 85],
       [71, 58, 75, 72, 69],
       [76, 97, 49, 63, 53]])
```

```
In [241]:
print("\nDouble bracket indexing\n----")
print("Element in row index 1 and column index 2:", mat[1][1])
mat
Double bracket indexing
Element in row index 1 and column index 2: 58
Out[241]:
array([[35, 75, 44, 59, 85],
      [71, 58, 75, 72, 69],
      [76, 97, 49, 63, 53]])
In [242]:
print("\nSingle bracket with comma indexing\n-----")
print("Element in row index 1 and column index 2:", mat[1,2])
print("\nRow or column extract\n----")
Single bracket with comma indexing
Element in row index 1 and column index 2: 75
Row or column extract
-----
In [172]:
print("Entire row at index 2:", mat[2])
print("Entire column at index 3:", mat[:,3])
Entire row at index 2: [22 10 57 59 61]
Entire column at index 3: [92 88 59]
In [7]:
print("\nSubsetting sub-matrices\n-----")
print("Matrix with row indices 1 and 2 and column indices 3 and 4\n", mat[1:3,3:5])
Subsetting sub-matrices
Matrix with row indices 1 and 2 and column indices 3 and 4
 [[14 32]
 [62 27]]
```

```
In [156]:

print("Matrix with row indices 0 and 1 and column indices 1 and 3\n", mat[0:2,[1,3]])

Matrix with row indices 0 and 1 and column indices 1 and 3

[[77 69]
[64 17]]

Subseting
```

```
In [173]:
                                                                                   M
mat = np.array(ri(10,100,15)).reshape(3,5)
print("Matrix of random 2-digit numbers\n-----\n",mat)
mat>50
Matrix of random 2-digit numbers
 [[51 99 67 32 43]
 [53 61 80 97 51]
 [82 76 18 26 11]]
Out[173]:
array([[ True, True, True, False, False],
      [ True, True, True, True],
      [ True, True, False, False, False]])
                                                                                   M
In [174]:
print ("Elements greater than 50\n", mat[mat>50])
Elements greater than 50
```

Slicing

[21 22 23] [31 32 33]]

```
In [175]:

mat = np.array([[11,12,13],[21,22,23],[31,32,33]])
print("Original matrix")
print(mat)

Original matrix
[[11 12 13]
```

[51 99 67 53 61 80 97 51 82 76]

```
In [176]:
                                                                                            H
mat_slice = mat[:2,:2]
print ("\nSliced matrix")
print(mat_slice)
print ("\nChange the sliced matrix")
Sliced matrix
[[11 12]
 [21 22]]
Change the sliced matrix
In [18]:
                                                                                            M
mat_slice[0,0] = 1000
print (mat_slice)
[[1000
         12]
         22]]
[ 21
                                                                                            H
In [75]:
print("\nBut the original matrix? WHOA! It got changed too!")
print(mat)
But the original matrix? WHOA! It got changed too!
[[11 12 13]
 [21 22 23]
 [31 32 33]]
In [177]:
# Little different way to create a copy of the slixed matrix
print ("\nDoing it again little differently now...\n")
mat = np.array([[11,12,13],[21,22,23],[31,32,33]])
print("Original matrix")
print(mat)
Doing it again little differently now...
Original matrix
[[11 12 13]
[21 22 23]
 [31 32 33]]
```

```
In [21]:
mat_slice = np.array(mat[:2,:2]) # Notice the np.array command to create a new array not ju
print ("\nSliced matrix")
print(mat_slice)
Sliced matrix
[[11 12]
 [21 22]]
                                                                                           H
In [178]:
print ("\nChange the sliced matrix")
mat_slice[0,0] = 1000
print (mat_slice)
Change the sliced matrix
[[1000
         12]
 [ 21
         22]]
In [22]:
                                                                                           M
print("\nBut the original matrix? NO CHANGE this time:)")
print(mat)
But the original matrix? NO CHANGE this time:)
[[11 12 13]
 [21 22 23]
 [31 32 33]]
```

Universal Functions

```
In [247]:
                                                                                H
mat1 = np.array(ri(1,10,9)).reshape(3,3)
mat2 = np.array(ri(1,10,9)).reshape(3,3)
print("\n1st Matrix of random single-digit numbers\n-----
print("\n2nd Matrix of random single-digit numbers\n------
1st Matrix of random single-digit numbers
 [[3 3 9]
 [6 1 8]
 [5 9 1]]
2nd Matrix of random single-digit numbers
 [[3 5 7]
 [4 4 9]
 [1 2 5]]
                                                                                M
In [248]:
mat1*mat2
Out[248]:
array([[ 9, 15, 63],
      [24, 4, 72],
      [5, 18, 5]])
In [ ]:
                                                                                H
In [249]:
#print("\nAddition\n----\n", mat1+mat2)
print("\nMultiplication\n-----\n", mat1@mat2)
Multiplication
 [[ 30 45 93]
 [ 30 50 91]
 [ 52 63 121]]
```

```
In [251]:
print("\nDivision\n----\n", mat1/0)
#print("\nLineaer combination: 3*A - 2*B\n-----
                                               ----\n", 3*mat1-2*mat2)
Division
[[inf inf inf]
[inf inf inf]
[inf inf inf]]
/Users/sudhanshukumar/anaconda3/lib/python3.7/site-packages/ipykernel_launch
er.py:1: RuntimeWarning: divide by zero encountered in true_divide
 """Entry point for launching an IPython kernel.
In [252]:
                                                                        M
print("\nAddition of a scalar (100)\n-----\n", 100+mat1)
Addition of a scalar (100)
[[103 103 109]
[106 101 108]
[105 109 101]]
In [253]:
print("\nExponentiation, matrix cubed here\n-----\n", ma
Exponentiation, matrix cubed here
[[ 27 27 729]
[216 1 512]
[125 729 1]]
Exponentiation, sq-root using pow function
[[ 27 27 729]
[216
      1 512]
[125 729 1]]
```

Broadcasting

```
In [174]: ▶
```

#NumPy operations are usually done on pairs of arrays on an element-by-element basis.
#In the simplest case, the two arrays must have exactly the same shape.
#NumPy's broadcasting rule relaxes this constraint when the arrays' shapes meet certain con
#When operating on two arrays, NumPy compares their shapes element-wise. It starts with the
#dimensions, and works its way forward. Two dimensions are compatible when
#they are equal, or one of them is 1

```
In [254]:
                                                                                            H
start = np.zeros((4,4))
start= start+100
start
Out[254]:
array([[100., 100., 100., 100.],
       [100., 100., 100., 100.],
       [100., 100., 100., 100.],
       [100., 100., 100., 100.]])
In [256]:
                                                                                            H
# create a rank 1 ndarray with 3 values
add_rows = np.array([1, 0, 2,5])
print(add_rows)
[1 0 2 5]
                                                                                            H
In [257]:
y = start + add_rows # add to each row of 'start' using broadcasting
print(y)
[[101. 100. 102. 105.]
 [101. 100. 102. 105.]
 [101. 100. 102. 105.]
 [101. 100. 102. 105.]]
In [262]:
                                                                                            H
# create an ndarray which is 4 x 1 to broadcast across columns
add_cols = np.array([[0,1,2,3]])
add_cols = add_cols.T
print(add_cols)
[[0]]
 [1]
 [2]
```

[3]]

```
In [263]:
                                                                                          H
# add to each column of 'start' using broadcasting
y = start + add_cols
print(y)
[[100. 100. 100. 100.]
 [101. 101. 101. 101.]
 [102. 102. 102. 102.]
 [103. 103. 103. 103.]]
In [264]:
                                                                                          H
# this will just broadcast in both dimensions
add_scalar = np.array([100])
print(start+y)
[[200. 200. 200. 200.]
 [201. 201. 201. 201.]
 [202. 202. 202. 202.]
 [203. 203. 203. 203.]]
Array Math
In [265]:
                                                                                          H
```

```
mat1 = np.array(ri(1,10,9)).reshape(3,3)
mat2 = np.array(ri(1,10,9)).reshape(3,3)
print("\n1st Matrix of random single-digit numbers\n\n",mat1)
print("\n2nd Matrix of random single-digit numbers\n----\n",mat2)
```

```
1st Matrix of random single-digit numbers
```

```
[[9 5 4]
[1 7 3]
[8 3 5]]
```

2nd Matrix of random single-digit numbers

```
-----
[[4 8 6]
[7 6 8]
[4 6 5]]
```

```
In [266]:
print("\nSq-root of 1st matrix using np\n----\n", np.sqrt(mat1))
Sq-root of 1st matrix using np
 [[3.
             2.23606798 2.
            2.64575131 1.73205081]
 [2.82842712 1.73205081 2.23606798]]
In [267]:
print("\nExponential power of 1st matrix using np\n",'-'*50,"\n", np.exp(mat1))
Exponential power of 1st matrix using np
 [[8.10308393e+03 1.48413159e+02 5.45981500e+01]
 [2.71828183e+00 1.09663316e+03 2.00855369e+01]
 [2.98095799e+03 2.00855369e+01 1.48413159e+02]]
In [268]:
                                                                                         M
print("\n10-base logarithm on 1st matrix using np\n",'-'*50,"\n", np.log10(mat1))
print(mat1)
print(mat2)
10-base logarithm on 1st matrix using np
 [[0.95424251 0.69897
                        0.60205999]
            0.84509804 0.47712125]
 [0.90308999 0.47712125 0.69897
                                 ]]
[[9 5 4]
[1 7 3]
[8 3 5]]
[[4 8 6]
 [7 6 8]
 [4 6 5]]
```

```
In [270]:
print("\nModulo reminder using np\n",'-'*50,"\n", np.fmod(mat1,mat2))
mat1%mat2
Modulo reminder using np
 [[1 5 4]
 [1 1 3]
 [0 3 0]]
Out[270]:
array([[1, 5, 4],
       [1, 1, 3],
       [0, 3, 0]])
In [2]:
print("\nCombination of functions by showing exponetial decay of a sine wave\n",'-'*70)
Combination of functions by showing exponetial decay of a sine wave
In [3]:
A = np.linspace(0,12*np.pi,1001)
In [4]:
                                                                                           H
Α
Out[4]:
                , 0.03769911, 0.07539822, ..., 37.62371362,
       37.66141273, 37.69911184])
```