

Python 101 - Python Libraries for Data Analysis - Numpy and Pandas

December 8, 2024

1 TASK #1: DEFINE SINGLE AND MULTI-DIMENSIONAL NUMPY ARRAYS

```
[1]: # NumPy is a Linear Algebra Library used for multidimensional arrays  
# NumPy brings the best of two worlds: (1) C/Fortran computational efficiency, ↵  
↪ (2) Python language easy syntax  
  
import numpy as np  
  
# Let's define a one-dimensional array  
list_1 = [50, 60, 80, 100, 200, 300, 500, 600]  
list_1
```

```
[1]: [50, 60, 80, 100, 200, 300, 500, 600]
```

```
[2]: # Let's create a numpy array from the list "my_list"  
my_numpy_array = np.array(list_1)  
my_numpy_array
```

```
[2]: array([ 50,  60,  80, 100, 200, 300, 500, 600])
```

```
[3]: type(my_numpy_array)
```

```
[3]: numpy.ndarray
```

```
[4]: # Multi-dimensional (Matrix definition)  
my_matrix = np.array([[2, 5, 8], [7,3,6]])  
my_matrix
```

```
[4]: array([[2, 5, 8],  
          [7, 3, 6]])
```

MINI CHALLENGE #1: - Write a code that creates the following 2x4 numpy array

```
[[3 7 9 3]  
 [4 3 2 2]]
```

```
[5]: challenge1 = np.array([[3, 7, 9, 3], [4, 3, 2, 2]])
challenge1
```

```
[5]: array([[3, 7, 9, 3],
           [4, 3, 2, 2]])
```

2 TASK #2: LEVERAGE NUMPY BUILT-IN METHODS AND FUNCTIONS

```
[6]: # "rand()" uniform distribution between 0 and 1
x = np.random.rand(20)
x
```

```
[6]: array([0.35935739, 0.43325741, 0.88238455, 0.31420639, 0.45434715,
           0.22535573, 0.48179596, 0.95975847, 0.03825129, 0.60871258,
           0.82256522, 0.11334626, 0.84976678, 0.41353044, 0.23532522,
           0.96353187, 0.36422946, 0.53042833, 0.45259285, 0.97201079])
```

```
[7]: # you can create a matrix of random number as well
x = np.random.rand(3,3)
x
```

```
[7]: array([[0.89305307, 0.30931093, 0.7313589 ],
           [0.2738033 , 0.68030488, 0.85975632],
           [0.60179017, 0.41297283, 0.87441933]])
```

```
[8]: # "randint" is used to generate random integers between upper and lower bounds
x = np.random.randint(1, 50)
x
```

```
[8]: 16
```

```
[9]: # "randint" can be used to generate a certain number of random itegers as
↳ follows
x = np.random.randint(1 ,100, 15)
x
```

```
[9]: array([23, 54, 72, 40, 22, 22, 60, 27, 62, 12, 82, 73, 49, 17, 95])
```

```
[10]: # np.arange creates an evenly spaced values within a given interval
x = np.arange(1,30)
x
```

```
[10]: array([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15, 16, 17,
           18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29])
```

```
[11]: # create a diagonal of ones and zeros everywhere else
x = np.eye(7)
x
```

```
[11]: array([[1., 0., 0., 0., 0., 0., 0.],
           [0., 1., 0., 0., 0., 0., 0.],
           [0., 0., 1., 0., 0., 0., 0.],
           [0., 0., 0., 1., 0., 0., 0.],
           [0., 0., 0., 0., 1., 0., 0.],
           [0., 0., 0., 0., 0., 1., 0.],
           [0., 0., 0., 0., 0., 0., 1.]])
```

```
[12]: # Matrix of ones
x = np.ones((7,7))
x
```

```
[12]: array([[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.]])
```

```
[13]: # Array of zeros
x = np.zeros(8)
x
```

```
[13]: array([0., 0., 0., 0., 0., 0., 0., 0.]])
```

MINI CHALLENGE #2: - Write a code that takes in a positive integer “x” from the user and creates a 1x10 array with random numbers ranging from 0 to “x”

```
[14]: X = int(input("Please enter a positive value"))
```

Please enter a positive value2

```
[15]: x = np.random.randint(1, X, 10)
x
```

```
[15]: array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1])
```

3 TASK #3: PERFORM MATHEMATICAL OPERATIONS IN NUMPY

```
[16]: # np.arange() returns an evenly spaced values within a given interval
```

```
[17]: x = np.arange(1, 10)
      x
```

```
[17]: array([1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
[18]: # Add 2 numpy arrays together
      y = np.arange(1, 10)
      y
```

```
[18]: array([1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
[19]: sum = x + y
      sum
```

```
[19]: array([ 2,  4,  6,  8, 10, 12, 14, 16, 18])
```

```
[20]: squared = x ** 2
      squared
```

```
[20]: array([ 1,  4,  9, 16, 25, 36, 49, 64, 81])
```

```
[21]: sqrt = np.sqrt(squared)
      sqrt
```

```
[21]: array([1., 2., 3., 4., 5., 6., 7., 8., 9.])
```

```
[22]: z = np.exp(y)
      z
```

```
[22]: array([2.71828183e+00, 7.38905610e+00, 2.00855369e+01, 5.45981500e+01,
          1.48413159e+02, 4.03428793e+02, 1.09663316e+03, 2.98095799e+03,
          8.10308393e+03])
```

MINI CHALLENGE #3: - Given the X and Y values below, obtain the distance between them

X = [5, 7, 20]

Y = [9, 15, 4]

```
[23]: X = np.array([5, 7, 20])
      Y = np.array([9, 15, 4])
      distance = np.sqrt(X ** 2 + Y ** 2)
      distance
```

```
[23]: array([10.29563014, 16.55294536, 20.39607805])
```

4 TASK #4: PERFORM ARRAYS SLICING AND INDEXING

```
[24]: my_numpy_array = np.array([3, 5, 6, 2, 8, 10, 20, 50])  
my_numpy_array
```

```
[24]: array([ 3,  5,  6,  2,  8, 10, 20, 50])
```

```
[25]: # Access specific index from the numpy array  
my_numpy_array[0]
```

```
[25]: 3
```

```
[26]: # Starting from the first index 0 up until and NOT including the last element  
my_numpy_array[0:len(my_numpy_array) - 1]
```

```
[26]: array([ 3,  5,  6,  2,  8, 10, 20])
```

```
[27]: # Broadcasting, altering several values in a numpy array at once  
my_numpy_array[0:4] = 7  
my_numpy_array
```

```
[27]: array([ 7,  7,  7,  7,  8, 10, 20, 50])
```

```
[28]: # Let's define a two dimensional numpy array  
matrix = np.random.randint(1, 10, (4, 4))  
matrix
```

```
[28]: array([[8, 4, 4, 2],  
          [7, 7, 8, 4],  
          [2, 4, 4, 9],  
          [2, 4, 1, 5]])
```

```
[29]: # Get a row from a matrix  
matrix[-1]
```

```
[29]: array([2, 4, 1, 5])
```

```
[30]: # Get one element  
matrix[0][2]
```

```
[30]: 4
```

MINI CHALLENGE #4: - In the following matrix, replace the last row with 0

X = [2 30 20 -2 -4]

```
[3 4 40 -3 -2]
[-3 4 -6 90 10]
[25 45 34 22 12]
[13 24 22 32 37]
```

```
[31]: X = np.array([[2, 30, 20, -2, -4],
                  [3, 4, 40, -3, -2],
                  [-3, 4, -6, 90, 10],
                  [25, 45, 34, 22, 12],
                  [13, 24, 22, 32, 37]])
X[-1] = 0
X
```

```
[31]: array([[ 2, 30, 20, -2, -4],
             [ 3,  4, 40, -3, -2],
             [-3,  4, -6, 90, 10],
             [25, 45, 34, 22, 12],
             [ 0,  0,  0,  0,  0]])
```

5 TASK #5: PERFORM ELEMENTS SELECTION (CONDITIONAL)

```
[32]: matrix = np.random.randint(1, 10, (5, 5))
matrix
```

```
[32]: array([[8, 3, 3, 9, 8],
             [7, 3, 7, 4, 4],
             [5, 4, 1, 9, 6],
             [6, 1, 9, 8, 2],
             [2, 9, 2, 8, 6]])
```

```
[33]: new_matrix = matrix[ matrix > 7 ]
new_matrix
```

```
[33]: array([8, 9, 8, 9, 9, 8, 9, 8])
```

```
[34]: # Obtain odd elements only
new_matrix = matrix[ matrix % 2 == 1]
new_matrix
```

```
[34]: array([3, 3, 9, 7, 3, 7, 5, 1, 9, 1, 9, 9])
```

MINI CHALLENGE #5: - In the following matrix, replace negative elements by 0 and replace odd elements with -2

```
X = [2 30 20 -2 -4]
     [3 4 40 -3 -2]
```

```

[-3  4 -6 90 10]
[25 45 34 22 12]
[13 24 22 32 37]

```

```

[35]: X = np.array([[2, 30, 20, -2, -4],
                  [3, 4, 40, -3, -2],
                  [-3, 4, -6, 90, 10],
                  [25, 45, 34, 22, 12],
                  [13, 24, 22, 32, 37]])

X[X < 0] = 0
X[X % 2 == 1] = -2
X

```

```

[35]: array([[ 2, 30, 20,  0,  0],
            [-2,  4, 40,  0,  0],
            [ 0,  4,  0, 90, 10],
            [-2, -2, 34, 22, 12],
            [-2, 24, 22, 32, -2]])

```

6 TASK #6: UNDERSTAND PANDAS FUNDAMENTALS

```

[36]: # Pandas is a data manipulation and analysis tool that is built on Numpy.
      # Pandas uses a data structure known as DataFrame (think of it as Microsoft
      ↪ excel in Python).
      # DataFrames empower programmers to store and manipulate data in a tabular
      ↪ fashion (rows and columns).
      # Series Vs. DataFrame? Series is considered a single column of a DataFrame.

```

```

[37]: import pandas as pd

```

```

[38]: # Let's define a two-dimensional Pandas DataFrame
      # Note that you can create a pandas dataframe from a python dictionary

bank_client_df = pd.DataFrame({'Bank Client ID': [111, 222, 333, 444],
                              'Bank Clien Name': ['Chanel', 'Steve', 'Mitch',
                              ↪ 'Ryan'],
                              'Net Worth [$]': [3500, 2900, 1000, 2000],
                              'Years with bank': [3, 4, 9, 5]})

bank_client_df

```

```

[38]:   Bank Client ID Bank Clien Name  Net Worth [$]  Years with bank
0           111         Chanel         3500           3
1           222         Steve         2900           4
2           333         Mitch         1000           9

```

3

444

Ryan

2000

5

```
[39]: # Let's obtain the data type
type(bank_client_df)
```

```
[39]: pandas.core.frame.DataFrame
```

```
[40]: # you can only view the first couple of rows using .head()
bank_client_df.head(2)
```

```
[40]:   Bank Client ID Bank Clie Name  Net Worth [$]  Years with bank
0           111          Chanel          3500           3
1           222          Steve          2900           4
```

```
[41]: # you can only view the last couple of rows using .tail()
bank_client_df.tail(2)
```

```
[41]:   Bank Client ID Bank Clie Name  Net Worth [$]  Years with bank
2           333          Mitch          1000           9
3           444          Ryan          2000           5
```

MINI CHALLENGE #6: - A portfolio contains a collection of securities such as stocks, bonds and ETFs. Define a dataframe named 'portfolio_df' that holds 3 different stock ticker symbols, number of shares, and price per share (feel free to choose any stocks) - Calculate the total value of the portfolio including all stocks

```
[42]: portfolio_df = pd.DataFrame({'Stock Symbols': ['MSFT', 'SP500', 'BTC'],
                                'Number of shares': [0.8400093, 6.976221, 0.007413],
                                'Price per Shares': [415.80, 113.42, 93724.53]})

Number_Shares = portfolio_df['Number of shares']
Price_Shares = portfolio_df['Price per Shares']

Price_Shares

Total = Number_Shares * Price_Shares
Total.sum()
```

```
[42]: 1835.2987936499999
```

7 TASK #7: PANDAS WITH CSV AND HTML DATA

```
[43]: # Pandas is used to read a csv file and store data in a DataFrame
import pandas as pd
house_price_df = pd.read_html('https://www.livingin-canada.com/
    house-prices-canada.html')
```



```
house_price_df[0]
```

```
[43]:
```

	City \
0	Vancouver, BC
1	Toronto, Ont
2	Ottawa, Ont
3	Calgary, Alb
4	Montreal, Que
5	Halifax, NS
6	Regina, Sask
7	Fredericton, NB

```
8 (adsbygoogle = window.adsbygoogle || []).push(...
```

	Average House Price \
0	\$1,036,000
1	\$870,000
2	\$479,000
3	\$410,000
4	\$435,000
5	\$331,000
6	\$254,000
7	\$198,000

```
8 (adsbygoogle = window.adsbygoogle || []).push(...
```

	12 Month Change
0	+ 2.63 %
1	+10.2 %
2	+ 15.4 %
3	- 1.5 %
4	+ 9.3 %
5	+ 3.6 %
6	- 3.9 %
7	- 4.3 %

```
8 (adsbygoogle = window.adsbygoogle || []).push(...
```

```
[44]: house_price_df[1]
```

```
[44]:
```

	Province \
0	British Columbia
1	Ontario
2	Alberta
3	Quebec
4	Manitoba
5	Saskatchewan
6	Nova Scotia
7	Prince Edward Island
8	Newfoundland / Labrador

```

9                 New Brunswick
10                Canadian Average
11  (adsbygoogle = window.adsbygoogle || []).push(...)

                Average House Price \
0                 $736,000
1                 $594,000
2                 $353,000
3                 $340,000
4                 $295,000
5                 $271,000
6                 $266,000
7                 $243,000
8                 $236,000
9                 $183,000
10                $488,000
11  (adsbygoogle = window.adsbygoogle || []).push(...)

                12 Month Change
0                 + 7.6 %
1                 - 3.2 %
2                 - 7.5 %
3                 + 7.6 %
4                 - 1.4 %
5                 - 3.8 %
6                 + 3.5 %
7                 + 3.0 %
8                 - 1.6 %
9                 - 2.2 %
10                - 1.3 %
11  (adsbygoogle = window.adsbygoogle || []).push(...)

```

```
[45]: # Read tabular data using read_html
```

```
[ ]:
```

```
[ ]:
```

MINI CHALLENGE #7: - Write a code that uses Pandas to read tabular US retirement data -
You can use data from here: <https://www.ssa.gov/oact/progdata/nra.html>

```
[46]: Us_retirement_df = pd.read_html('https://www.ssa.gov/oact/progdata/nra.html')
      Us_retirement_df[0]
```

```
[46]:                Year of birth \
0                1937 and prior
1                1938
```

```

2                                     1939
3                                     1940
4                                     1941
5                                     1942
6                                     1943-54
7                                     1955
8                                     1956
9                                     1957
10                                    1958
11                                    1959
12                                    1960 and later
13 Notes: 1. Persons born on January 1 of any yea...

```

```

                                     Age
0                                     65
1          65 and 2 months
2          65 and 4 months
3          65 and 6 months
4          65 and 8 months
5          65 and 10 months
6                                     66
7          66 and 2 months
8          66 and 4 months
9          66 and 6 months
10         66 and 8 months
11         66 and 10 months
12                                     67
13 Notes: 1. Persons born on January 1 of any yea...

```

8 TASK #8: PANDAS OPERATIONS

```

[47]: # Let's define a dataframe as follows:
bank_client_df = pd.DataFrame({'Bank Client ID': [ 111, 222, 333, 444],
                               'Bank Clien Name': ['Chanel', 'Steve', 'Mitch', 'Ryan'],
                               'Net Worth [$]': [3500, 29000, 10000, 2000],
                               'Years with bank': [3, 4, 9, 5]})

bank_client_df

```

```

[47]:   Bank Client ID Bank Clien Name  Net Worth [$]  Years with bank
0           111         Chanel         3500           3
1           222          Steve        29000           4
2           333          Mitch        10000           9
3           444           Ryan         2000           5

```

```
[48]: # Pick certain rows that satisfy a certain criteria
df_loyal = bank_client_df[ bank_client_df['Years with bank'] >= 5]
df_loyal
```

```
[48]:   Bank Client ID Bank Client Name  Net Worth [$]  Years with bank
2             333           Mitch         10000             9
3             444           Ryan          2000             5
```

```
[49]: # Delete a column from a DataFrame
del bank_client_df['Bank Client ID']
bank_client_df
```

```
[49]:   Bank Client Name  Net Worth [$]  Years with bank
0           Chanel         3500             3
1           Steve        29000             4
2           Mitch        10000             9
3           Ryan         2000             5
```

MINI CHALLENGE #8: - Using “bank_client_df” DataFrame, leverage pandas operations to only select high network individuals with minimum \$5000 - What is the combined network for all customers with 5000+ network?

```
[51]: df_network = bank_client_df[ bank_client_df['Net Worth [$]' ] >= 5000]
df_network

total_network = df_network['Net Worth [$]'].sum()
total_network
```

```
[51]: 39000
```

9 TASK #9: PANDAS WITH FUNCTIONS

```
[ ]: # Let's define a dataframe as follows:
bank_client_df = pd.DataFrame({'Bank client ID':[111, 222, 333, 444],
                               'Bank Client Name':['Chanel', 'Steve', 'Mitch',
                               ↪ 'Ryan'],
                               'Net worth [$]':[3500, 29000, 10000, 2000],
                               'Years with bank':[3, 4, 9, 5]})

bank_client_df
```

```
[ ]: # Define a function that increases all clients network (stocks) by a fixed
↪ value of 20% (for simplicity sake)
def increase_networth(balance):
    return balance * 1.2
```

```
[ ]: # You can apply a function to the DataFrame
bank_client_df['Net Worth [$]'].apply(increase_networth)
```

```
[ ]: bank_client_df['Bank Client Name'].apply(len)
```

MINI CHALLENGE #9: - Define a function that triples the stock prices and adds \$200 - Apply the function to the DataFrame - Calculate the updated total network of all clients combined

```
[ ]: def triple_stock(value):
      return value * 3.0 + 200

triple_stock = bank_client_df['Net worth [$]'].apply(triple_stock)
triple_stock.sum()
```

10 TASK #10: PERFORM SORTING AND ORDERING IN PANDAS

```
[52]: # Let's define a dataframe as follows:
bank_client_df = pd.DataFrame({'Bank client ID':[111, 222, 333, 444],
                               'Bank Client Name':['Chanel', 'Steve', 'Mitch',
                               ↪'Ryan'],
                               'Net worth [$]':[3500, 29000, 10000, 2000],
                               'Years with bank':[3, 4, 9, 5]})

bank_client_df
```

```
[52]:
```

	Bank client ID	Bank Client Name	Net worth [\$]	Years with bank
0	111	Chanel	3500	3
1	222	Steve	29000	4
2	333	Mitch	10000	9
3	444	Ryan	2000	5

```
[53]: # You can sort the values in the dataframe according to number of years with
      ↪bank
bank_client_df.sort_values(by = 'Years with bank')
```

```
[53]:
```

	Bank client ID	Bank Client Name	Net worth [\$]	Years with bank
0	111	Chanel	3500	3
1	222	Steve	29000	4
3	444	Ryan	2000	5
2	333	Mitch	10000	9

```
[54]: # Note that nothing changed in memory! you have to make sure that inplace is
      ↪set to True
bank_client_df
```

```
[54]: Bank client ID Bank Client Name Net worth [$] Years with bank
0      111      Chanel      3500      3
1      222      Steve      29000      4
2      333      Mitch      10000      9
3      444      Ryan       2000      5
```

```
[55]: # Set inplace = True to ensure that change has taken place in memory
bank_client_df.sort_values(by = 'Years with bank', inplace = True)
```

```
[56]: # Note that now the change (ordering) took place
bank_client_df
```

```
[56]: Bank client ID Bank Client Name Net worth [$] Years with bank
0      111      Chanel      3500      3
1      222      Steve      29000      4
3      444      Ryan       2000      5
2      333      Mitch      10000      9
```

11 TASK #11: PERFORM CONCATENATING AND MERGING WITH PANDAS

```
[ ]: # Check this out: https://pandas.pydata.org/pandas-docs/stable/user\_guide/merging.html
```

```
[57]: df1 = pd.DataFrame({ 'A': ['A0', 'A1', 'A2', 'A3'],
                        'B': ['B0', 'B1', 'B2', 'B3'],
                        'C': ['C0', 'C1', 'C2', 'C3'],
                        'D': ['D0', 'D1', 'D2', 'D3']},
                        index = [0,1,2,3])
```

```
[58]: df1
```

```
[58]:   A  B  C  D
0  A0 B0 C0 D0
1  A1 B1 C1 D1
2  A2 B2 C2 D2
3  A3 B3 C3 D3
```

```
[59]: df2 = pd.DataFrame({ 'A': ['A4', 'A5', 'A6', 'A7'],
                        'B': ['B4', 'B5', 'B6', 'B7'],
                        'C': ['C4', 'C5', 'C6', 'C7'],
                        'D': ['D4', 'D5', 'D6', 'D7']},
                        index = [4,5,6,7])
```

```
[60]: df2
```

```
[60]:
```

	A	B	C	D
0	A4	B4	C4	D4
1	A5	B5	C5	D5
2	A6	B6	C6	D6
3	A7	B7	C7	D7

```
[62]: df3 = pd.DataFrame({ 'A': ['A8', 'A9', 'A10', 'A11'],
                           'B': ['B8', 'B9', 'B10', 'B11'],
                           'C': ['C8', 'C9', 'C10', 'C11'],
                           'D': ['D8', 'D9', 'D10', 'D11']},
                           index = [8,9,10,11])
```

```
[63]: df3
```

```
[63]:
```

	A	B	C	D
8	A8	B8	C8	D8
9	A9	B9	C9	D9
10	A10	B10	C10	D10
11	A11	B11	C11	D11

```
[65]: pd.concat([df1, df2, df3])
```

```
[65]:
```

	A	B	C	D
0	A0	B0	C0	D0
1	A1	B1	C1	D1
2	A2	B2	C2	D2
3	A3	B3	C3	D3
0	A4	B4	C4	D4
1	A5	B5	C5	D5
2	A6	B6	C6	D6
3	A7	B7	C7	D7
8	A8	B8	C8	D8
9	A9	B9	C9	D9
10	A10	B10	C10	D10
11	A11	B11	C11	D11

12 TASK #12: PROJECT AND CONCLUDING REMARKS

- Define a dataframe named 'Bank_df_1' that contains the first and last names for 5 bank clients with IDs = 1, 2, 3, 4, 5
- Assume that the bank got 5 new clients, define another dataframe named 'Bank_df_2' that contains a new clients with IDs = 6, 7, 8, 9, 10
- Let's assume we obtained additional information (Annual Salary) about all our bank customers (10 customers)
- Concatenate both 'bank_df_1' and 'bank_df_2' dataframes
- Merge client names and their newly added salary information using the 'Bank Client ID'
- Let's assume that you became a new client to the bank

- Define a new DataFrame that contains your information such as client ID (choose 11), first name, last name, and annual salary.
- Add this new dataframe to the original dataframe 'bank_df_all'.

```
[125]: bank_df_1 = pd.DataFrame({'Bank Client ID': [1,2,4,5],
                                'First Name': ['Emily', 'Alex','Sarah','Michael'],
                                'Last Name': ['Thompson', 'Johson',
                                ↪'Robinson','Ngueyn']})
```

```
[128]: bank_df_2 = pd.DataFrame({
    'Bank Client ID': [6, 7, 8, 9, 10],
    'First Name': ['Daniel', 'Sophia', 'James', 'Olivia', 'Liam'],
    'Last Name': ['Smith', 'Brown', 'Taylor', 'Garcia', 'Clark']
})

raw_data = {'Bank Client ID': [1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
            'Annual Salary': [2500, 3500, 4500, 4800, 3200, 3300, 3400, 2300,
            ↪2200, 35000]}

bank_df_salary = pd.DataFrame(raw_data, columns = ['Bank Client ID', 'Annual
            ↪Salary'])
bank_df_salary
```

```
[128]:
```

	Bank Client ID	Annual Salary
0	1	2500
1	2	3500
2	3	4500
3	4	4800
4	5	3200
5	6	3300
6	7	3400
7	8	2300
8	9	2200
9	10	35000

```
[129]: bank_df_all = pd.concat([bank_df_1, bank_df_2])
bank_df_all
```

```
[129]:
```

	Bank Client ID	First Name	Last Name
0	1	Emily	Thompson
1	2	Alex	Johson
2	4	Sarah	Robinson
3	5	Michael	Ngueyn
0	6	Daniel	Smith
1	7	Sophia	Brown
2	8	James	Taylor
3	9	Olivia	Garcia

4 10 Liam Clark

```
[130]: Bank_df_all = pd.merge(bank_df_all, bank_df_salary, on = 'Bank Client ID')
Bank_df_all
```

```
[130]:
```

	Bank Client ID	First Name	Last Name	Annual Salary
0	1	Emily	Thompson	2500
1	2	Alex	Johson	3500
2	4	Sarah	Robinson	4800
3	5	Michael	Ngueyn	3200
4	6	Daniel	Smith	3300
5	7	Sophia	Brown	3400
6	8	James	Taylor	2300
7	9	Olivia	Garcia	2200
8	10	Liam	Clark	35000

```
[131]: new_client = pd.DataFrame({'Bank Client ID': [11],
                                'First Name': ['Kevin'],
                                'Last Name': ['Truong'],
                                'Annual Salary': [1000]})

bank_df_all = pd.concat([Bank_df_all, new_client], ignore_index = True)
bank_df_all
```

```
[131]:
```

	Bank Client ID	First Name	Last Name	Annual Salary
0	1	Emily	Thompson	2500
1	2	Alex	Johson	3500
2	4	Sarah	Robinson	4800
3	5	Michael	Ngueyn	3200
4	6	Daniel	Smith	3300
5	7	Sophia	Brown	3400
6	8	James	Taylor	2300
7	9	Olivia	Garcia	2200
8	10	Liam	Clark	35000
9	11	Kevin	Truong	1000

13 EXCELLENT JOB!

14 MINI CHALLENGES SOLUTIONS

MINI CHALLENGE #1 SOLUTION: - Write a code that creates the following 2x4 numpy array

```
[[3 7 9 3]
 [4 3 2 2]]
```

```
[ ]: x = np.array([[3, 7, 9, 3] , [4, 3, 2, 2]])
x
```

MINI CHALLENGE #2 SOLUTION: - Write a code that takes in a positive integer “x” from the user and creates a 1x10 array with random numbers ranging from 0 to “x”

```
[ ]: x = int(input("Please enter a positive integer value: "))
x = np.random.randint(1, x, 10)
x
```

```
[ ]:
```

MINI CHALLENGE #3 SOLUTION: - Given the X and Y values below, obtain the distance between them

```
X = [5, 7, 20]
Y = [9, 15, 4]
```

```
[ ]: X = np.array([5, 7, 20])
Y = np.array([9, 15, 4])
Z = np.sqrt(X**2 + Y**2)
Z
```

MINI CHALLENGE #4 SOLUTION: - In the following matrix, replace the last row with 0

```
X = [2 30 20 -2 -4]
     [3 4 40 -3 -2]
     [-3 4 -6 90 10]
     [25 45 34 22 12]
     [13 24 22 32 37]
```

```
[ ]: X = np.array([[2, 30, 20, -2, -4],
                  [3, 4, 40, -3, -2],
                  [-3, 4, -6, 90, 10],
                  [25, 45, 34, 22, 12],
                  [13, 24, 22, 32, 37]])
```

```
[ ]: X[4] = 0
X
```

MINI CHALLENGE #5 SOLUTION: - In the following matrix, replace negative elements by 0 and replace odd elements with -2

```
X = [2 30 20 -2 -4]
     [3 4 40 -3 -2]
     [-3 4 -6 90 10]
     [25 45 34 22 12]
     [13 24 22 32 37]
```

```
[ ]: X = np.array([[2, 30, 20, -2, -4],
                  [3, 4, 40, -3, -2],
                  [-3, 4, -6, 90, 10],
                  [25, 45, 34, 22, 12],
                  [13, 24, 22, 32, 37]])
```

```
[13, 24, 22, 32, 37]])

X[X<0] = 0
X[X%2==1] = -2
X
```

MINI CHALLENGE #6 SOLUTION: - A portfolio contains a collection of securities such as stocks, bonds and ETFs. Define a dataframe named 'portfolio_df' that holds 3 different stock ticker symbols, number of shares, and price per share (feel free to choose any stocks) - Calculate the total value of the portfolio including all stocks

```
[ ]: portfolio_df = pd.DataFrame({'stock ticker symbols':['AAPL', 'AMZN', 'T'],
                                'price per share [$]':[3500, 200, 40],
                                'Number of stocks':[3, 4, 9]})

portfolio_df
```

```
[ ]: stocks_dollar_value = portfolio_df['price per share [$]'] *
    ↪ portfolio_df['Number of stocks']
print(stocks_dollar_value)
print('Total portfolio value = {}'.format(stocks_dollar_value.sum()))
```

MINI CHALLENGE #7 SOLUTION: - Write a code that uses Pandas to read tabular US retirement data - You can use data from here: <https://www.ssa.gov/oact/progdata/nra.html>

```
[ ]: # Read tabular data using read_html
retirement_age_df = pd.read_html('https://www.ssa.gov/oact/progdata/nra.html')
retirement_age_df
```

MINI CHALLENGE #8 SOLUTION: - Using "bank_client_df" DataFrame, leverage pandas operations to only select high networth individuals with minimum \$5000 - What is the combined networth for all customers with 5000+ network?

```
[ ]: df_high_networth = bank_client_df[ (bank_client_df['Net worth [$]'] >= 5000) ]
df_high_networth
```

```
[ ]: df_high_networth['Net worth [$]'].sum()
```

MINI CHALLENGE #9 SOLUTION: - Define a function that triples the stock prices and adds \$200 - Apply the function to the DataFrame - Calculate the updated total network of all clients combined

```
[ ]: def networth_update(balance):
    return balance * 3 + 200
```

```
[ ]: # You can apply a function to the DataFrame
results = bank_client_df['Net worth [$]'].apply(networth_update)
results
```

```
[ ]: results.sum()
```

PROJECT SOLUTION:

```
[ ]: # Creating a dataframe from a dictionary
# Let's define a dataframe with a list of bank clients with IDs = 1, 2, 3, 4, 5

raw_data = {'Bank Client ID': ['1', '2', '3', '4', '5'],
            'First Name': ['Nancy', 'Alex', 'Shep', 'Max', 'Allen'],
            'Last Name': ['Rob', 'Ali', 'George', 'Mitch', 'Steve']}

Bank_df_1 = pd.DataFrame(raw_data, columns = ['Bank Client ID', 'First Name', 'Last Name'])
Bank_df_1

# Let's define another dataframe for a separate list of clients (IDs = 6, 7, 8, 9, 10)
raw_data = {
    'Bank Client ID': ['6', '7', '8', '9', '10'],
    'First Name': ['Bill', 'Dina', 'Sarah', 'Heather', 'Holly'],
    'Last Name': ['Christian', 'Mo', 'Steve', 'Bob', 'Michelle']}
Bank_df_2 = pd.DataFrame(raw_data, columns = ['Bank Client ID', 'First Name', 'Last Name'])
Bank_df_2

# Let's assume we obtained additional information (Annual Salary) about our bank customers
# Note that data obtained is for all clients with IDs 1 to 10
raw_data = {
    'Bank Client ID': ['1', '2', '3', '4', '5', '6', '7', '8', '9', '10'],
    'Annual Salary [$ /year]': [25000, 35000, 45000, 48000, 49000, 32000, 33000, 34000, 23000, 22000]}
bank_df_salary = pd.DataFrame(raw_data, columns = ['Bank Client ID', 'Annual Salary [$ /year]'])
bank_df_salary

# Let's concatenate both dataframes #1 and #2
# Note that we now have client IDs from 1 to 10
bank_df_all = pd.concat([Bank_df_1, Bank_df_2])
bank_df_all

# Let's merge all data on 'Bank Client ID'
bank_df_all = pd.merge(bank_df_all, bank_df_salary, on = 'Bank Client ID')
```

```
bank_df_all
```

```
[ ]: new_client = {  
    'Bank Client ID': ['11'],  
    'First Name': ['Ry'],  
    'Last Name': ['Aly'],  
    'Annual Salary [$ /year]' : [1000]}  
new_client_df = pd.DataFrame(new_client, columns = ['Bank Client ID', 'First_↵  
    ↵Name', 'Last Name', 'Annual Salary [$ /year]'])  
new_client_df
```

```
[ ]: new_df = pd.concat([bank_df_all, new_client_df], axis = 0)  
new_df
```