

67070501021 นายธนบุรย์ ธิแก้ว

Lab 1: Basic Python Programming

CPE232 Data Models

[1] Variable

1.1 Number Variable

```
In [1]: num = 100 #integer variable
         num2 = 12.5 #float variable
         print(num)
         print(num2)

         print(num + num2)    #addition
         print(num - num2)    #subtraction
         print(num * num2)    #multiplication
         print( num / num2)   #division
```

```
100
12.5
112.5
87.5
1250.0
8.0
```

1.2 String Variable

```
In [2]: string = "Data Models"
         print(string) #print complete string

         print("Hello " + string)      #print concatenated string

         # slicing string
         print(string[0])            #print first character of the string
         print(string[:4])           #print first to 4th character of the string
         print(string[5:])           #print 6th to last character of the string
         print(string[1:4])           #print 2nd to 4th character of the string
         print(string * 2)            #print string 2 times
```

```
Data Models
Hello Data Models
D
Data
Models
ata
Data ModelsData Models
```

```
In [3]: # format string
code = "CPE232"

print(f"{code}: Data Models")
```

```
CPE232: Data Models
```

1.3 Boolean Variable

```
In [4]: #boolean variable
boolean = True
boolean2 = False

print(boolean)           #print boolean variable
print(not boolean)      #print opposite of boolean variable
print(boolean and boolean2) #print boolean and boolean2
print(boolean or boolean2) #print boolean or boolean2
```

```
True
False
False
True
```

1.4 List Variable

```
In [5]: #List variable
list = ["Data", 20, 123.23, 40, 50]
another_list = ["Models", 60]

print(list)               #print complete list
print(list[0])            #print first element of the list
print(list[1:3])          #print 2nd to 3rd element of the list
print(list[2:])            #print 3rd to last element of the list
print(another_list)        #print complete another_list
print(another_list * 2)    #print another_list two times
print(list + another_list) #print concatenated list

list[0] = "CPE232"         #change first element of the list
print(list)                #print complete list

# add element at the end
list.append(True)
print(list)

# pop element by index
list.pop(1)
print(list)
```

```
[ 'Data', 20, 123.23, 40, 50]
Data
[20, 123.23]
[123.23, 40, 50]
['Models', 60]
['Models', 60, 'Models', 60]
['Data', 20, 123.23, 40, 50, 'Models', 60]
['CPE232', 20, 123.23, 40, 50]
['CPE232', 20, 123.23, 40, 50, True]
['CPE232', 123.23, 40, 50, True]
```

1.5 Tuple Variable

```
In [6]: #tuple variable
tuple = ("Data",20,123.23,40,50)
another_tuple = ("Models",60)

print(tuple)                      #print complete tuple
print(tuple[0])                   #print first element of the tuple
print(tuple[1:3])                 #print 2nd to 3rd element of the tuple
print(tuple[2:])                  #print 3rd to Last element of the tuple
print(tuple * 2)                  #print tuple two times
print(tuple + another_tuple)       #print concatenated tuple

('Data', 20, 123.23, 40, 50)
Data
(20, 123.23)
(123.23, 40, 50)
('Data', 20, 123.23, 40, 50, 'Data', 20, 123.23, 40, 50)
('Data', 20, 123.23, 40, 50, 'Models', 60)
```

```
In [7]: tuple[0] = "CPE232"      # tuple is immutable dtype, we can't change its value
```

```
-----
```

```
TypeError                                     Traceback (most recent call last)
Cell In[7], line 1
----> 1 tuple[0] = "CPE232"                  # tuple is immutable dtype, we can't change its
value after declared

TypeError: 'tuple' object does not support item assignment
```

1.6 Dictionary Variable

```
In [8]: #dictionary variable
dictionary = {"name":"Alice","age":21}
another_dictionary = {}
another_dictionary["name"] = "Bob"
another_dictionary["age"] = 21

print(dictionary)                      #print complete dictionary
print(dictionary["name"])            #print value for specific key
print(dictionary.keys())              #print all the keys
print(dictionary.values())             #print all the values
```

```

print(dictionary.items())           #print all the items
print(another_dictionary)          #print complete another_dictionary

{'name': 'Alice', 'age': 21}
Alice
dict_keys(['name', 'age'])
dict_values(['Alice', 21])
dict_items([('name', 'Alice'), ('age', 21)])
{'name': 'Bob', 'age': 21}

```

1.7 Set Variable

```

In [9]: #set variable
my_set = {"Data", 20, 123.23, 60, "Data"}
another_set = {"Models", 60}

print(my_set)                      #print complete set (duplicates are removed)

#set operations
print(my_set.union(another_set))    # print all unique elements from both sets
print(my_set.intersection(another_set)) # print common elements from both sets

# add element
my_set.add("New Item")
my_set.add(20)
print(my_set)                      # try adding a duplicate again

# remove element
my_set.remove(123.23)
print(my_set)

{123.23, 'Data', 20, 60}
{'Models', 'Data', 20, 123.23, 60}
{60}
{'New Item', 'Data', 20, 123.23, 60}
{'New Item', 'Data', 20, 60}

```

[2] Control Flow

2.1 IF ... ELIF ... ELSE

```

In [10]: number = 123
number2 = 34

if number > number2:
    print("number is greater than number2")
elif number < number2:
    print("number is less than number2")
else:
    print("number is equal to number2")

```

number is greater than number2

2.2 try ... except

```
In [11]: number1 = "number"
number2 = 2

try:
    number1 + number2
except:                                     #error handling
    print("What are you doing?")
```

```
What are you doing?
```

[3] Loop

3.1 For Loop

```
In [12]: #for Loops
for num in range(0,10):
    print(num)
```

```
0
1
2
3
4
5
6
7
8
9
```

```
In [13]: #for Loops with step
for num in range(5,15,2):
    print(num)
```

```
5
7
9
11
13
```

```
In [14]: #for Loop with enumerate

list = ["Alice","Bob","Charlie","Daisy"]

for index, name in enumerate(list):
    print(f"{index}: {name}")
```

```
0: Alice
1: Bob
2: Charlie
3: Daisy
```

```
In [15]: #for Loop with List
```

```
for name in list:  
    print(name)
```

```
Alice  
Bob  
Charlie  
Daisy
```

```
In [16]: #continue in for Loop
```

```
list = [1,23,7,"hello",True,1123,43,23,12]  
  
for element in list:  
    if type(element) != int:  
        continue  
    print(element)
```

```
1  
23  
7  
1123  
43  
23  
12
```

```
In [17]: #break in for Loop
```

```
list = [1,23,7,"hello",True,1123,43,23,12]  
  
for element in list:  
    if type(element) != int:  
        break  
    print(element)
```

```
1  
23  
7
```

3.2 While loop

```
In [18]: #while Loop
```

```
list = ["Alice","Bob","Charlie","Daisy"]  
count = 0  
  
while count < len(list):  
    print(list[count])  
    count += 1
```

```
Alice  
Bob  
Charlie  
Daisy
```

```
In [19]: #continue in while Loop
```

```
list = [1,23,7,"hello",True,1123,43,23,12]
count = 0

while count < len(list):
    if type(list[count]) != int:
        count += 1
        continue
    print(list[count])
    count += 1
```

```
1
23
7
1123
43
23
12
```

```
In [20]: #break in while Loop
```

```
list = [1,23,7,"hello",True,1123,43,23,12]
count = 0

while count < len(list):
    if type(list[count]) != int:
        break
    print(list[count])
    count += 1
```

```
1
23
7
```

[4] Function

```
In [21]: #define function
```

```
def function_name (arg1, arg2):
    return arg1 + arg2

#calling function
function_name(1,2)
```

```
Out[21]: 3
```

```
In [22]: #define function with default argument
```

```
def function_with_default_arg(arg1, arg2 = 10, arg3 = 20 , arg4 = 30):
    return arg1 + arg2 + arg3 + arg4

result_1 = function_with_default_arg(1)
result_2 = function_with_default_arg(1,2,5)
result_3 = function_with_default_arg(1,2,5,10)
```

```
print(result_1)
print(result_2)
print(result_3)
```

```
61
38
18
```

```
In [23]: #multiple arguments
def function_with_multiple_arg(*args):
    print(args)
    print(type(args))
    sum = 0
    for num in args:
        sum += num

    return sum

function_with_multiple_arg(1,2,3,4,5)
```

```
(1, 2, 3, 4, 5)
<class 'tuple'>
```

```
Out[23]: 15
```

```
In [24]: #Lambda function
lambda_function = lambda arg1, arg2: arg1 + arg2

print(lambda_function(1,2))
```

```
3
```

[5] OOP

```
In [25]: class Car:  
    """A simple attempt to represent a car."""  
  
    def __init__(self, make, model, year):  
        """Initialize attributes to describe a car."""  
        self.make = make  
        self.model = model  
        self.year = year  
        self.odometer_reading = 0  
  
    def get_description_name(self):  
        """Return a neatly formatted descriptive name."""  
        long_name = f"{self.year} {self.make} {self.model}"  
        return long_name.title()  
  
    def read_odometer(self):  
        """Print a statement showing the car's mileage."""  
        print(f"This car has {self.odometer_reading} miles on it.")  
  
    def update_odometer(self, mileage):  
        """Set the odometer reading to the given value."""  
        self.odometer_reading = mileage
```

```
In [26]: my_car = Car("Honda", "Civic", 2016)  
my_car.read_odometer()  
  
# update odometer  
my_car.update_odometer(50)  
my_car.read_odometer()
```

This car has 0 miles on it.
This car has 50 miles on it.

[6] File Handling

6.1 Text File

```
In [27]: # write file  
with open("test.txt","w") as file:  
    file.write("Hello World")  
  
# read file  
with open("test.txt","r") as file:  
    print(file.read())
```

Hello World

6.2 CSV File

```
In [28]: import csv  
  
with open("test.csv", "w", newline='') as file:  
    writer = csv.writer(file)
```

```
writer.writerow(["Name", "Surname"])
writer.writerow(["Alice", "Johnson"])
writer.writerow(["Bob", "Smith"])
```

```
In [29]: import csv

with open("test.csv", "r") as file:
    reader = csv.reader(file)
    for row in reader:
        print(row)

['Name', 'Surname']
['Alice', 'Johnson']
['Bob', 'Smith']
```

6.3 JSON file

```
In [30]: import json

data = {
    "username": "admin",
    "password": "admin",
    "role": "sleeping",
}

with open('test.json', 'w') as f:
    json.dump(data, f)
```

```
In [31]: with open("test.json", "r") as f:
    data = json.load(f)

data
```

```
Out[31]: {'username': 'admin', 'password': 'admin', 'role': 'sleeping'}
```

[7] Libraries

7.1 Numpy

import numpy library

```
In [32]: import numpy as np
```

ndarray initialization

Construct using python list

```
In [33]: # 1d ndarray from 1d python list
list_a1=[1,2,3.5]
```

```
arr_a1=np.array(list_a1)
arr_a1
```

```
Out[33]: array([1. , 2. , 3.5])
```

```
In [34]: # 2d ndarray from 2d python List (list of list)
list_a2=[[1,2],[3,4],[5,6]]
arr_a2=np.array(list_a2)
arr_a2
```

```
Out[34]: array([[1, 2],
 [3, 4],
 [5, 6]])
```

```
In [35]: list_a3=[ [[1,2],[2,3]], [[3,4],[4,5]] ]
arr_a3=np.array(list_a3)
arr_a3
```

```
Out[35]: array([[[1, 2],
 [2, 3]],
 [[3, 4],
 [4, 5]])
```

or construct using some numpy classes and functions

```
In [36]: np.zeros(5)
```

```
Out[36]: array([0., 0., 0., 0., 0.])
```

```
In [37]: np.ones((3,4),dtype=float)
```

```
Out[37]: array([[1., 1., 1., 1.],
 [1., 1., 1., 1.],
 [1., 1., 1., 1.]])
```

```
In [38]: np.eye(3, 3)
```

```
Out[38]: array([[1., 0., 0.],
 [0., 1., 0.],
 [0., 0., 1.]])
```

```
In [39]: np.full((4,),999)
```

```
Out[39]: array([999, 999, 999, 999])
```

```
In [40]: np.arange(3,10,2)
```

```
Out[40]: array([3, 5, 7, 9])
```

```
In [41]: np.linspace(0, 1, 5)
```

```
Out[41]: array([0. , 0.25, 0.5 , 0.75, 1. ])
```

```
In [42]: np.random.choice(['a', 'b'], 9)
```

```
Out[42]: array(['a', 'b', 'b', 'a', 'b', 'a', 'b', 'b', 'a'], dtype='<U1')
```

```
In [43]: np.random.randn(10)
```

```
Out[43]: array([ 1.35591984, -0.91629096,  1.58048896,  1.18457964, -1.28496854,
   0.30538524,  1.74836835,  0.41450679, -0.35972875,  0.2995627 ])
```

ndarray properties

```
In [44]: list_a=[[1,2,3,4],[5,6,7,8],[9,10,11,12]]
arr_a=np.array(list_a)
arr_a
```

```
Out[44]: array([[ 1,  2,  3,  4],
   [ 5,  6,  7,  8],
   [ 9, 10, 11, 12]])
```

```
In [45]: print("dimension:", arr_a.ndim)
print("shape:", arr_a.shape)
print("data type:", arr_a.dtype)
print("array's size (elements)", arr_a.size)
```

```
dimension: 2
shape: (3, 4)
data type: int64
array's size (elements) 12
```

Reshaping & Modification

from this original ndarray

```
In [46]: arr_a
```

```
Out[46]: array([[ 1,  2,  3,  4],
   [ 5,  6,  7,  8],
   [ 9, 10, 11, 12]])
```

try to convert into 3D array

```
In [47]: arr_a.reshape((2,2,3))
```

```
Out[47]: array([[[ 1,  2,  3],
   [ 4,  5,  6]],
  [[ 7,  8,  9],
   [10, 11, 12]]])
```

sometimes you may resize for same dimension where only known some dimension, insert -1 for unknown len

```
In [48]: arr_a.reshape((-1,6))
```

```
Out[48]: array([[ 1,  2,  3,  4,  5,  6],
   [ 7,  8,  9, 10, 11, 12]])
```

Would you like to try this?

```
In [49]: arr_a.reshape((-1,5))
```

```
-----  
ValueError                                     Traceback (most recent call last)  
Cell In[49], line 1  
----> 1 arr_a.reshape((-1,5))  
  
ValueError: cannot reshape array of size 12 into shape (5)
```

[Q1] From the above cell, explain in your own words why it worked or did not work.

Ans: It did not work because array in shape 5 can be formed by Array with elements which is multiplicant of 5 (5,10,15,...) but arr_a have 12 elements

Next, try to append any value(s) into existing 2darray

```
In [50]: np.append(arr_a, 13)
```

```
Out[50]: array([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13])
```

```
In [51]: np.append(arr_a, arr_a[0])
```

```
Out[51]: array([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12,  1,  2,  3,  4])
```

```
In [52]: np.append(arr_a, arr_a[0].reshape((1,-1)),axis=0)
```

```
Out[52]: array([[ 1,  2,  3,  4],
   [ 5,  6,  7,  8],
   [ 9, 10, 11, 12],
   [ 1,  2,  3,  4]])
```

```
In [53]: np.append(arr_a, arr_a[:,0].reshape((-1,1)),axis=1)
```

```
Out[53]: array([[ 1,  2,  3,  4,  1],
   [ 5,  6,  7,  8,  5],
   [ 9, 10, 11, 12,  9]])
```

```
In [54]: np.concatenate([arr_a,arr_a])
```

```
Out[54]: array([[ 1,  2,  3,  4],
   [ 5,  6,  7,  8],
   [ 9, 10, 11, 12],
   [ 1,  2,  3,  4],
   [ 5,  6,  7,  8],
   [ 9, 10, 11, 12]])
```

```
In [55]: np.concatenate([arr_a,arr_a],axis=1)
```

```
Out[55]: array([[ 1,  2,  3,  4,  1,  2,  3,  4],
   [ 5,  6,  7,  8,  5,  6,  7,  8],
   [ 9, 10, 11, 12,  9, 10, 11, 12]])
```

indexing & slicing

from this original array again

```
In [56]: arr_a
```

```
Out[56]: array([[ 1,  2,  3,  4],
   [ 5,  6,  7,  8],
   [ 9, 10, 11, 12]])
```

try to access all element at the first row

```
In [57]: arr_a[1]
```

```
Out[57]: array([5, 6, 7, 8])
```

then you would like to access the second element from the first row

```
In [58]: arr_a[1][2]
arr_a[1,2]
```

```
Out[58]: np.int64(7)
```

Next, try to access all element start from the 2nd element in the first row

```
In [59]: arr_a[1,1:]
```

```
Out[59]: array([6, 7, 8])
```

```
In [60]: arr_a[:2, 1:]
```

```
Out[60]: array([[2, 3, 4],
   [6, 7, 8]])
```

sometimes you may specify some row number using list within indexing

```
In [61]: arr_a[[1,2,1], 1:]
```

```
Out[61]: array([[ 6,  7,  8],
   [10, 11, 12],
   [ 6,  7,  8]])
```

Boolean slicing

based on this original array

```
In [62]: arr_a
```

```
Out[62]: array([[ 1,  2,  3,  4],  
                 [ 5,  6,  7,  8],  
                 [ 9, 10, 11, 12]])
```

try to filter all elements which more than 5

```
In [63]: arr_a>5
```

```
Out[63]: array([[False, False, False, False],  
                  [False, True, True, True],  
                  [ True, True, True, True]])
```

Next, try to filter all elements which more than 5 and less than 10

```
In [64]: (arr_a>5)&(arr_a<10)
```

```
Out[64]: array([[False, False, False, False],  
                  [False, True, True, True],  
                  [ True, False, False, False]])
```

Run the cell below and answer a question.

```
In [65]: arr_a[(arr_a>5)&(arr_a<10)]
```

```
Out[65]: array([6, 7, 8, 9])
```

[Q2] From the above cell, explain in your own words how the output came about?

Ans: The output came from elements in arr_a which is intersection between arr_a > 5 (6,7,8,9,10,11,12) and arr_a < 10 (1,2,3,4,5,6,7,8,9) so the intersection is (6,7,8,9) or index 5 to index 8 and we filter arr_a using the intersection index

Try running the cell below.

```
In [119... arr_a[(arr_a>5) and (arr_a<10)]
```

```
-----  
ValueError  
Cell In[119], line 1  
----> 1 arr_a[(arr_a>5) and (arr_a<10)]
```

Traceback (most recent call last)

```
ValueError: The truth value of an array with more than one element is ambiguous. Use  
a.any() or a.all()
```

[Q3] Explain in your own words why the above cell gives an error.

Ans: "and" operator is using on single element in this case arr_a > 5 have 12 elements which is array([[False, False, False, False], [False, True, True, True], [True, True, True, True]]) so "and" operator don't know what is Truth value of arr_a > 5 and occurs the error

while "&" operator check element by element(bitwise) so it works on arr_a > 5 & arr_a < 10 both have same 12 elements it intersect it one by one so it can give the right output

[Q4] And what should be written instead so that the code is error-free?

Ans: arr_a[(arr_a>5).any() and (arr_a<10).any()]

```
In [120]: arr_a[(arr_a>5).any() and (arr_a<10).any()]
```

```
Out[120]: array([[1, 2, 3, 4],
                  [5, 6, 7, 8],
                  [9, 10, 11, 12]])
```

Basic operations

```
In [67]: list_b=[[1,2,3,4],[1,2,3,4],[1,2,3,4]]
arr_b=np.array(list_b)
arr_b
```

```
Out[67]: array([[1, 2, 3, 4],
                 [1, 2, 3, 4],
                 [1, 2, 3, 4]])
```

This is some operations for only 1 array

```
In [68]: np.sqrt(arr_b)
```

```
Out[68]: array([[1.        , 1.41421356, 1.73205081, 2.        ],
                 [1.        , 1.41421356, 1.73205081, 2.        ],
                 [1.        , 1.41421356, 1.73205081, 2.        ]])
```

This is some operations for 2 arrays with the same shape

```
In [69]: arr_a-arr_b
```

```
Out[69]: array([[0, 0, 0, 0],
                 [4, 4, 4, 4],
                 [8, 8, 8, 8]])
```

```
In [70]: np.add(arr_a, arr_b)
```

```
Out[70]: array([[ 2,  4,  6,  8],
                 [ 6,  8, 10, 12],
                 [10, 12, 14, 16]])
```

Next, try to operate with 1 array and one numeric variable

```
In [71]: arr_a*3
```

```
Out[71]: array([[ 3,  6,  9, 12],  
                 [15, 18, 21, 24],  
                 [27, 30, 33, 36]])
```

```
In [72]: 1+arr_a**2
```

```
Out[72]: array([[ 2,  5, 10, 17],  
                  [26, 37, 50, 65],  
                  [82, 101, 122, 145]])
```

Broadcasting

```
In [73]: arr_c=np.array([1,2,3])  
arr_d=np.array([[3],[5],[8]])
```

```
In [74]: arr_c-arr_d
```

```
Out[74]: array([[-2, -1,  0],  
                  [-4, -3, -2],  
                  [-7, -6, -5]])
```

Basic aggregations

```
In [75]: arr_a  
  
        print(arr_a)  
        print("sum:", arr_a.sum())  
        print("mean:", arr_a.mean())  
        print("min:", arr_a.min())  
        print("max:", arr_a.max())  
        print("standard deviation:", arr_a.std())  
  
[[ 1  2  3  4]  
 [ 5  6  7  8]  
 [ 9 10 11 12]]  
sum: 78  
mean: 6.5  
min: 1  
max: 12  
standard deviation: 3.452052529534663
```

ndarray axis

```
In [76]: # column axis  
arr_a.sum(axis=0)
```

```
Out[76]: array([15, 18, 21, 24])
```

```
In [77]: # row axis  
arr_a.sum(axis=1)
```

```
Out[77]: array([10, 26, 42])
```

[Q5] Summarize the value of the argument *axis*, what is the value for row-wise summation and column-wise summation, respectively?

Ans: row-wise summation axis = 1 and column-wise summation axis = 0

7.2 Pandas

Series

```
In [84]: import pandas as pd  
import numpy as np
```

```
In [85]: pd.Series(np.random.randn(6))
```

```
Out[85]: 0    -0.890746  
1    -0.391122  
2    -0.844234  
3    -0.615343  
4     0.572527  
5     0.880289  
dtype: float64
```

```
In [86]: pd.Series(np.random.randn(6), index=['a', 'b', 'c', 'd', 'e', 'f'])
```

```
Out[86]: a    -0.593951  
b    -0.670971  
c     0.217029  
d    -0.504141  
e     0.010253  
f    -1.194180  
dtype: float64
```

Constructing Dataframe

Constructing DataFrame from a dictionary

```
In [81]: d = {'col1':[1,2], 'col2': [3,4]}
```

```
In [87]: df = pd.DataFrame(data=d)  
df
```

```
Out[87]:   col1  col2  
0      1      3  
1      2      4
```

```
In [88]: d2 = {'Name':['Joe','Nat','Harry','Sam','Monica'],  
          'Age': [20,21,19,20,22]}
```

```
df2 = pd.DataFrame(data=d2)
df2
```

Out[88]:

	Name	Age
0	Joe	20
1	Nat	21
2	Harry	19
3	Sam	20
4	Monica	22

Constructing DataFrame from a List

```
In [89]: marks_list = [85.10, 77.80, 91.54, 88.78, 60.55]
df3 = pd.DataFrame(marks_list, columns=['Marks'])
df3
```

Out[89]:

	Marks
0	85.10
1	77.80
2	91.54
3	88.78
4	60.55

Creating DataFrame from file

```
In [90]: # Read csv file from path and store to df for create dataframe
df = pd.read_csv('nss15.csv')
df
```

Out[90]:

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis
0	150733174	7/11/2015	15.7762	V	5	Male	NaN	57
1	150734723	7/6/2015	83.2157	S	36	Male	White	57
2	150817487	8/2/2015	74.8813	L	20	Female	NaN	71
3	150717776	6/26/2015	15.7762	V	61	Male	NaN	71
4	150721694	7/4/2015	74.8813	L	88	Female	Other	62
...
334834	150739278	5/31/2015	15.0591	V	7	Male	NaN	59
334835	150733393	7/11/2015	5.6748	C	3	Female	Black	68
334836	150819286	7/24/2015	15.7762	V	38	Male	NaN	71
334837	150823002	8/8/2015	97.9239	M	38	Female	White	59
334838	150723074	6/20/2015	49.2646	M	5	Female	White	57

334839 rows × 12 columns



Viewing DataFrame information

(.shape, .head, .tail, .info, select column, .unique, .describe, select row with .loc and .iloc)

Check simple information

In [91]: # Check dimension by .shape
df.shape

Out[91]: (334839, 12)

In [92]: # Display the first 5 rows by default
df.head()

Out[92]:

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	body
0	150733174	7/11/2015	15.7762	V	5	Male	NaN		57
1	150734723	7/6/2015	83.2157	S	36	Male	White		57
2	150817487	8/2/2015	74.8813	L	20	Female	NaN		71
3	150717776	6/26/2015	15.7762	V	61	Male	NaN		71
4	150721694	7/4/2015	74.8813	L	88	Female	Other		62



```
In [93]: # Display the first 3 rows  
df.head(3)
```

```
Out[93]:
```

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	body
0	150733174	7/11/2015	15.7762	V	5	Male	NaN		57
1	150734723	7/6/2015	83.2157	S	36	Male	White		57
2	150817487	8/2/2015	74.8813	L	20	Female	NaN		71



```
In [94]: # Display the last 5 rows by default  
df.tail()
```

```
Out[94]:
```

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	
334834	150739278	5/31/2015	15.0591	V	7	Male	NaN		59
334835	150733393	7/11/2015	5.6748	C	3	Female	Black		68
334836	150819286	7/24/2015	15.7762	V	38	Male	NaN		71
334837	150823002	8/8/2015	97.9239	M	38	Female	White		59
334838	150723074	6/20/2015	49.2646	M	5	Female	White		57



```
In [95]: # Overview information of dataframe  
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 334839 entries, 0 to 334838  
Data columns (total 12 columns):  
 #   Column           Non-Null Count  Dtype     
---  --  
 0   caseNumber      334839 non-null  int64    
 1   treatmentDate   334839 non-null  object    
 2   statWeight      334839 non-null  float64   
 3   stratum          334839 non-null  object    
 4   age              334839 non-null  int64    
 5   sex              334837 non-null  object    
 6   race             205014 non-null  object    
 7   diagnosis        334839 non-null  int64    
 8   bodyPart         334839 non-null  int64    
 9   disposition      334839 non-null  int64    
 10  location         334839 non-null  int64    
 11  product          334839 non-null  int64    
 dtypes: float64(1), int64(7), object(4)  
memory usage: 30.7+ MB
```

Select column, multiple column, with condition

```
In [96]: df.columns
```

```
Out[96]: Index(['caseNumber', 'treatmentDate', 'statWeight', 'stratum', 'age', 'sex',  
                 'race', 'diagnosis', 'bodyPart', 'disposition', 'location', 'product'],  
                 dtype='object')
```

```
In [97]: # column slicing  
df['age']
```

```
Out[97]: 0      5  
1      36  
2      20  
3      61  
4      88  
      ..  
334834    7  
334835    3  
334836   38  
334837   38  
334838    5  
Name: age, Length: 334839, dtype: int64
```

```
In [98]: df.age
```

```
Out[98]: 0      5  
1      36  
2      20  
3      61  
4      88  
      ..  
334834    7  
334835    3  
334836   38  
334837   38  
334838    5  
Name: age, Length: 334839, dtype: int64
```

Viewing the unique value

```
In [99]: df.race.unique()
```

```
Out[99]: array([nan, 'White', 'Other', 'Black', 'Asian', 'American Indian'],  
                 dtype=object)
```

Null values

```
In [100...]: df.isnull().sum()
```

```
Out[100...]: caseNumber      0  
treatmentDate      0  
statWeight        0  
stratum           0  
age                0  
sex                2  
race              129825  
diagnosis         0  
bodyPart          0  
disposition       0  
location          0  
product           0  
dtype: int64
```

Describe

```
In [101...]: df['age'].describe()
```

```
Out[101...]: count    334839.000000  
mean      31.385451  
std       26.105098  
min       0.000000  
25%      10.000000  
50%      23.000000  
75%      51.000000  
max      107.000000  
Name: age, dtype: float64
```

Slicing dataframe

```
In [102...]: # select multiple column  
df[['treatmentDate', 'statWeight', 'age', 'sex']]
```

Out[102...]

	treatmentDate	statWeight	age	sex
0	7/11/2015	15.7762	5	Male
1	7/6/2015	83.2157	36	Male
2	8/2/2015	74.8813	20	Female
3	6/26/2015	15.7762	61	Male
4	7/4/2015	74.8813	88	Female
...
334834	5/31/2015	15.0591	7	Male
334835	7/11/2015	5.6748	3	Female
334836	7/24/2015	15.7762	38	Male
334837	8/8/2015	97.9239	38	Female
334838	6/20/2015	49.2646	5	Female

334839 rows × 4 columns

In [103...]

```
#select by condition
df[df['sex'] == 'Male']
```

Out[103...]

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	b
0	150733174	7/11/2015	15.7762	V	5	Male	NaN		57
1	150734723	7/6/2015	83.2157	S	36	Male	White		57
3	150717776	6/26/2015	15.7762	V	61	Male	NaN		71
6	150713483	6/8/2015	15.7762	V	25	Male	Black		51
7	150704114	6/14/2015	83.2157	S	53	Male	White		57
...
334824	150607827	5/27/2015	5.6748	C	1	Male	White		71
334825	150600190	5/28/2015	80.8381	S	5	Male	NaN		56
334833	150747217	7/24/2015	83.2157	S	2	Male	NaN		62
334834	150739278	5/31/2015	15.0591	V	7	Male	NaN		59
334836	150819286	7/24/2015	15.7762	V	38	Male	NaN		71

182501 rows × 12 columns



In [104...]

```
# select by multiple condition
df[(df['sex'] == 'Male') & (df['age'] > 80)]
```

Out[104...]

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	b
8	150736558	7/16/2015	83.2157	S	98	Male	Black		59
63	150418623	1/12/2015	15.0591	V	97	Male	Other		62
97	150700375	6/28/2015	83.2157	S	85	Male	NaN		59
131	150940801	9/14/2015	15.7762	V	96	Male	NaN		62
177	160110774	12/19/2015	85.7374	S	81	Male	White		59
...
334616	160104368	12/30/2015	74.8813	L	86	Male	Other		71
334677	151115099	11/4/2015	16.5650	V	83	Male	NaN		63
334699	150633387	5/29/2015	74.8813	L	84	Male	NaN		53
334701	150515945	4/27/2015	97.9239	M	86	Male	NaN		57
334785	150733286	7/11/2015	15.7762	V	86	Male	White		71

6379 rows × 12 columns



Select row with .iloc

In [105...]

```
# row slicing
df.iloc[10:15]
```

Out[105...]

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bod
10	150734952	7/4/2015	15.7762	V	20	Male	Black		59
11	150821622	7/20/2015	83.2157	S	20	Female	White		57
12	150713631	7/4/2015	15.7762	V	11	Male	NaN		60
13	150666343	6/27/2015	15.7762	V	26	Female	White		62
14	150748843	7/16/2015	37.6645	L	33	Male	Asian		53



In [106...]

```
# column slicing
df.iloc[:,[0,1,2,3,4]]
```

Out[106...]

	caseNumber	treatmentDate	statWeight	stratum	age
0	150733174	7/11/2015	15.7762	V	5
1	150734723	7/6/2015	83.2157	S	36
2	150817487	8/2/2015	74.8813	L	20
3	150717776	6/26/2015	15.7762	V	61
4	150721694	7/4/2015	74.8813	L	88
...
334834	150739278	5/31/2015	15.0591	V	7
334835	150733393	7/11/2015	5.6748	C	3
334836	150819286	7/24/2015	15.7762	V	38
334837	150823002	8/8/2015	97.9239	M	38
334838	150723074	6/20/2015	49.2646	M	5

334839 rows × 5 columns

Select column and row with .loc

In [107...]

```
# select column and row by .Loc
df.loc[:, 'treatmentDate':'diagnosis']
```

Out[107...]

	treatmentDate	statWeight	stratum	age	sex	race	diagnosis
0	7/11/2015	15.7762	V	5	Male	NaN	57
1	7/6/2015	83.2157	S	36	Male	White	57
2	8/2/2015	74.8813	L	20	Female	NaN	71
3	6/26/2015	15.7762	V	61	Male	NaN	71
4	7/4/2015	74.8813	L	88	Female	Other	62
5	7/2/2015	5.6748	C	1	Female	White	71
6	6/8/2015	15.7762	V	25	Male	Black	51

In [108...]

```
# select row by condition
df.loc[df['age'] > 80, ['treatmentDate', 'age']]
```

```
Out[108...]
```

	treatmentDate	age
4	7/4/2015	88
8	7/16/2015	98
39	5/3/2015	88
46	4/15/2015	91
63	1/12/2015	97
...
334701	4/27/2015	86
334784	7/7/2015	82
334785	7/11/2015	86
334815	10/28/2015	85
334819	1/13/2015	85

20422 rows × 2 columns

[Q6] What is the difference between .iloc and .loc?

Ans: .iloc is index-based query while .loc is label-based query

for example df.loc[:6,'treatmentDate':'diagnosis'] equals to df.iloc[:7,1:8] loc using 'treatmentDate':'diagnosis' label to query but iloc using 1:8 to query

also loc can query using condition such as df.loc[df['age']>80, ['treatmentDate', 'age']]

```
In [121...]
```

```
df.loc[:6, 'treatmentDate':'diagnosis']
```

```
Out[121...]
```

	treatmentDate	statWeight	stratum	age	sex	race	diagnosis
0	7/11/2015	15.7762	V	5	Male	NaN	57
1	7/6/2015	83.2157	S	36	Male	White	57
2	8/2/2015	74.8813	L	20	Female	NaN	71
3	6/26/2015	15.7762	V	61	Male	NaN	71
4	7/4/2015	74.8813	L	88	Female	Other	62
5	7/2/2015	5.6748	C	1	Female	White	71
6	6/8/2015	15.7762	V	25	Male	Black	51

```
In [125...]
```

```
df.iloc[:7,1:8]
```

Out[125...]

	treatmentDate	statWeight	stratum	age	sex	race	diagnosis
0	7/11/2015	15.7762	V	5	Male	NaN	57
1	7/6/2015	83.2157	S	36	Male	White	57
2	8/2/2015	74.8813	L	20	Female	NaN	71
3	6/26/2015	15.7762	V	61	Male	NaN	71
4	7/4/2015	74.8813	L	88	Female	Other	62
5	7/2/2015	5.6748	C	1	Female	White	71
6	6/8/2015	15.7762	V	25	Male	Black	51

Basic aggregations

In [109...]

```
# count elements in column via pandas Series method
df['sex'].value_counts()
```

Out[109...]

sex	N
Male	182501
Female	152336
Name: count, dtype: int64	

In [110...]

```
# count elements in column via pandas function
pd.crosstab(df['sex'], columns='N')
```

Out[110...]

col_0	N
sex	
Female	152336
Male	182501

In [111...]

```
summary_by_sex = df.groupby('sex').agg({
    'age': 'mean',
    'statWeight': 'sum',
    'caseNumber': 'count',
    'stratum': lambda x: x.mode()[0]
}).rename(columns={'caseNumber': 'total_cases'})

print(summary_by_sex)
```

sex	age	statWeight	total_cases	stratum
Female	35.585699	6.208773e+06	152336	V
Male	27.879792	6.964776e+06	182501	V

In [112...]

```
df['age'].apply('mean')
```

Out[112...]

```
np.float64(31.38545091820248)
```

```
In [133...]: df['race'].fillna("").apply(len)
```

```
Out[133...]: 0      0
 1      5
 2      0
 3      0
 4      5
 ..
334834    0
334835    5
334836    0
334837    5
334838    5
Name: race, Length: 334839, dtype: int64
```

```
In [ ]: df['race'].apply(len) #data have NaN so len(Nan) is error because python treat data
```

```
-----  
TypeError                                         Traceback (most recent call last)  
Cell In[132], line 1  
----> 1 df[      ].apply(len)  
  
File c:\Users\win25\Desktop\Desktop\work\CPE232 Data models\w1\.venv\Lib\site-packages\pandas\core\series.py:4943, in Series.apply(self, func, convert_dtype, args, by_row, **kwargs)  
    4808 def apply(  
    4809     self,  
    4810     func: AggFuncType,  
    (...) 4815     **kwargs,  
    4816 ) -> DataFrame | Series:  
    4817     """  
    4818     Invoke function on values of Series.  
    4819  
    (...) 4934     dtype: float64  
    4935     """  
    4936     return SeriesApply(  
    4937         self,  
    4938         func,  
    4939         convert_dtype=convert_dtype,  
    4940         by_row=by_row,  
    4941         args=args,  
    4942         kwargs=kwargs,  
-> 4943     ).apply()  
  
File c:\Users\win25\Desktop\Desktop\work\CPE232 Data models\w1\.venv\Lib\site-packages\pandas\core\apply.py:1422, in SeriesApply.apply(self)  
    1419     return self.apply_compat()  
    1421 # self.func is Callable  
-> 1422 return self.apply_standard()  
  
File c:\Users\win25\Desktop\Desktop\work\CPE232 Data models\w1\.venv\Lib\site-packages\pandas\core\apply.py:1502, in SeriesApply.apply_standard(self)  
    1496 # row-wise access  
    1497 # apply doesn't have a `na_action` keyword and for backward compat reasons  
    1498 # we need to give `na_action="ignore"` for categorical data.  
    1499 # TODO: remove the `na_action="ignore"` when that default has been changed in  
n  
    1500 # Categorical (GH51645).  
    1501 action = "ignore" if isinstance(obj.dtype, CategoricalDtype) else None  
-> 1502 mapped = obj._map_values(  
    1503     mapper=curried, na_action=action, convert=self.convert_dtype  
    1504 )  
    1506 if len(mapped) and isinstance(mapped[0], ABCSeries):  
    1507     # GH#43986 Need to do list(mapped) in order to get treated as nested  
    1508     # See also GH#25959 regarding EA support  
    1509     return obj._constructor_expanddim(list(mapped), index=obj.index)  
  
File c:\Users\win25\Desktop\Desktop\work\CPE232 Data models\w1\.venv\Lib\site-packages\pandas\core\base.py:925, in IndexOpsMixin._map_values(self, mapper, na_action, convert)  
    922 if isinstance(arr, ExtensionArray):  
    923     return arr.map(mapper, na_action=na_action)  
--> 925 return algorithms.map_array(arr, mapper, na_action=na_action, convert=convert)
```

```
t)
```

```
File c:\Users\win25\Desktop\Desktop\work\CPE232 Data models\w1\.venv\Lib\site-packages\pandas\core\algorithms.py:1743, in map_array(arr, mapper, na_action, convert)
    1741 values = arr.astype(object, copy=False)
    1742 if na_action is None:
-> 1743     return lib.map_infer(values, mapper, convert=convert)
    1744 else:
    1745     return lib.map_infer_mask(
    1746         values, mapper, mask=isna(values).view(np.uint8), convert=convert
    1747     )

File pandas\_libs/lib.pyx:2999, in pandas._libs.lib.map_infer()

TypeError: object of type 'float' has no len()
```

[Q7] In the above cell, Explain why pandas .agg()/.apply() method accepts 'len' and 'lambda' without quotation marks.

Ans: Because len and lambda is function its mean we pass the function to make pandas calling it themself for each row

7.3 Matplotlib

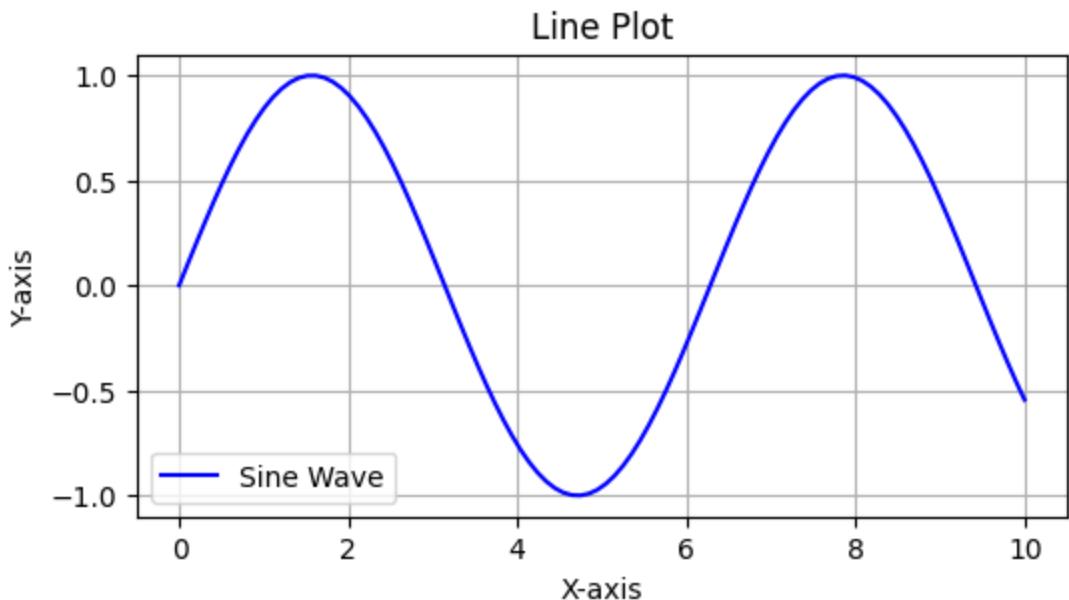
```
In [115...]  
import matplotlib.pyplot as plt  
import numpy as np
```

```
In [116...]  
x = np.linspace(0, 10, 100)  
y = np.sin(x)

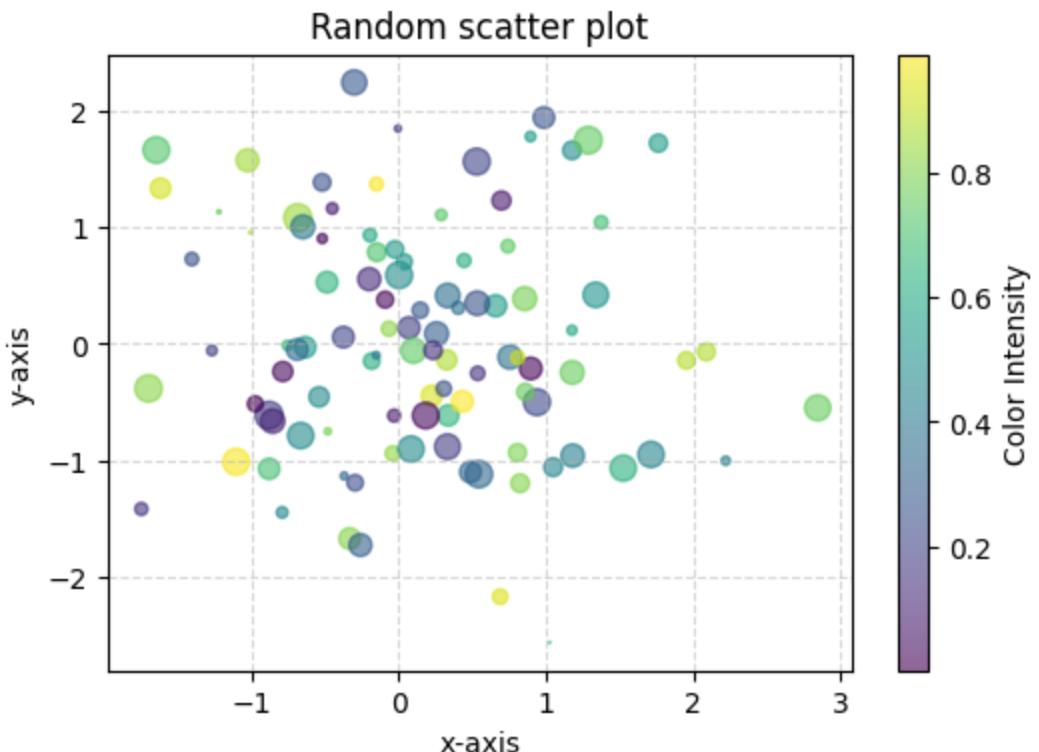
# Line plot

plt.figure(figsize=(6, 3))
plt.plot(x, y, label='Sine Wave', color='blue')
plt.title('Line Plot')
plt.xlabel('X-axis')
plt.ylabel('Y-axis')
plt.legend()
plt.grid(True)

plt.savefig("sine.png")
plt.show()
```



```
In [117]:  
x = np.random.randn(100)  
y = np.random.randn(100)  
colors = np.random.rand(100) # Random colors  
sizes = np.random.rand(100) * 100 # Random sizes  
  
# Scatter plot  
plt.figure(figsize=(6, 4))  
scatter = plt.scatter(x, y, c=colors, s=sizes, alpha=0.6, cmap='viridis')  
  
# Add styling and labels  
plt.title('Random scatter plot')  
plt.xlabel('x-axis')  
plt.ylabel('y-axis')  
plt.colorbar(scatter, label='Color Intensity')  
plt.grid(True, linestyle='--', alpha=0.5)  
  
# Show the plot  
plt.show()
```



```
In [118]: plt.figure(figsize=(5, 3))

# bar plot

df['sex'].value_counts().plot(kind='bar', color=['salmon', 'lightblue'])
plt.title('Gender Distribution')
plt.xlabel('Sex')
plt.ylabel('Count')
plt.xticks(rotation=0)
```

```
Out[118]: (array([0, 1]), [Text(0, 0, 'Male'), Text(1, 0, 'Female')])
```

