Lab 1: Basic Python Programming

CPE232 Data Models

Owner: 66070501043 - Phoorin Chinphuad

[1] Variable

→ 1.1 Number Variable

```
1 num = 100 #integer variable
2 num2 = 12.5 #float variable
3 print(num)
4 print(num2)
6 print(num + num2)
                     #addition
7 print(num - num2)
                     #subtraction
8 print(num * num2) #multiplication
9 print( num / num2) #division
→ 100
   12.5
   112.5
   87.5
   1250.0
   8.0
```

→ 1.2 String Variable

```
1 #string variable
 2 string = "Data Models"
 3 print(string) #print complete string
5 print("Hello " + string) #print concatenated string
6 print(string[0])
                             #print first character of the string
                           #print first to 4th character of the string
7 print(string[:4])
8 print(string[5:])
                            #print 6th to last character of the string
9 print(string[1:4])
                             #print 2nd to 4th character of the string
10 print(string * 2)
                              #print string 2 time
11
   Data Models
    Hello Data Models
    Data
    Models
    Data ModelsData Models
```

→ 1.3 Boolean Variable

✓ 1.4 List Variable

```
1 #list variable
 2 list = ["Data",20,123.23,40,50]
 3 another_list = ["Models",60]
 5 print(list)
                                 #print complete list
                                 #print first element of the list
 6 print(list[0])
 7 print(list[1:3])
                                 #print 2nd to 3rd element of the list
 8 print(list[2:])
                                  #print 3rd to last element of the list
 9 print(another_list)
                                  #print complete another_list
10 print(another_list * 2)
                                  #print another_list two times
11 print(list + another_list) #print concatenated list
12
13 list[0] = "CPE232"
                                  #change first element of the list
14 print(list)
                                  #print complete list
15
    ['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]

→ 1.5 Tuple Variable

 1 #tuple variable
 2 tuple = ("Data", 20, 123.23, 40, 50)
 3 another_tuple = ("Models",60)
 5 print(tuple)
                                      #print complete tuple
 6 print(tuple[0])
                                      #print first element of the tuple
 7 print(tuple[1:3])
                                      #print 2nd to 3rd element of the tuple
 8 print(tuple[2:])
                                      #print 3rd to last element of the tuple
 9 print(tuple * 2)
                                      #print tuple two times
10 print(tuple + another_tuple)
                                     #print concatenated tuple
11
    ('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)
 1 tuple[0] = "CPE232"
                                   #trying to change first element of the tuple but it cannot be changed so it gives error

→ 1.6 Dictionary Variable

 1 #dictionary variable
 2 dictionary = {"name":"Alice", "age":21}
 3 another_dictionary = {}
 4 another_dictionary["name"] = "Bob"
 5 another_dictionary["age"] = 21
 7 print(dictionary)
                                         #print complete dictionary
 8 print(dictionary["name"])
                                          #print value for specific key
 9 print(dictionary.keys())
                                          #print all the keys
10 print(dictionary.values())
                                         #print all the values
11 print(dictionary.items())
                                         #print all the items
12 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}
```

[2] Control Flow

```
1 \text{ number} = 123
 2 \text{ number2} = 34
 4 if number > number2:
      print("number is greater thanu number2")
 6 elif number < number2:
    print("number is less than number2")
 8 else:
      print("number is equal to number2")
10
number is greater thanu number2
[3] Loop
∨ 3.1 For Loop
 1 #for loops
 2 for num in range(0,10):
     print(num)
\overline{\pm}
   0
    4
 1 #for loop with list
 3 list = ["Alice","Bob","Charlie","Daisy"]
 4
 5 for name in list:
    print(name)
→ Alice
    Bob
    Charlie
    Daisy
 1 #continue in for loop
 3 list = [1,23,7,"hello",True,1123,43,23,12]
 4
 5 for element in list:
       if type(element) != int:
 6
 7
           continue
 8
       print(element)
    23
    1123
    43
    23
    12
 1 #break in for loop
 3 list = [1,23,7,"hello",True,1123,43,23,12]
 5 for element in list:
 6
      if type(element) != int:
           break
 8
       print(element)
```

```
1
23
7
```

√ 3.2 While loop

```
1 #while loop
 3 list = ["Alice", "Bob", "Charlie", "Daisy"]
 4 count = 0
 6 while count < len(list):
     print(list[count])
     count += 1
⇒ Alice
    Bob
    Charlie
    Daisy
 1 #continue in while loop
 3 list = [1,23,7,"hello",True,1123,43,23,12]
 4 count = 0
 5
 6 while count < len(list):</pre>
      if type(list[count]) != int:
 8
           count += 1
 9
           continue
10
      print(list[count])
      count += 1
11
    23
    1123
    23
 1 #break in while loop
 3 list = [1,23,7,"hello",True,1123,43,23,12]
 4 count = 0
 5
 6 while count < len(list):</pre>
      if type(list[count]) != int:
 8
           break
 9
      print(list[count])
10
      count += 1
    23
[4] Function
 1 #define function
 2 def function_name (arg1, arg2):
 3
      return arg1 + arg2
 5 #calling function
 6 function_name(1,2)
<del>_____</del> 3
 1 #define function with default argument
 2 def function_with_default_arg(arg1, arg2 = 10, arg3 = 20 , arg4 = 30):
 3
     return arg1 + arg2 + arg3 + arg4
 5 result_1 = function_with_default_arg(1)
 6 result_2 = function_with_default_arg(1,2,5)
```

```
7 result_3 = function_with_default_arg(1,2,5,10)
 9 print(result_1)
10 print(result_2)
11 print(result_3)
→ 61
     38
     18
 1 #multiple agument
 2 def function_with_multiple_arg(*args):
 3
       print(args)
 4
       print(type(args))
 5
       sum = 0
 6
       for num in args:
           sum += num
 7
 8
 9
       return sum
10
11 function_with_multiple_arg(1,2,3,4,5)
1 #lambda function
 2 lambda_function = lambda arg1, arg2: arg1 + arg2
 4 print(lambda_function(1,2))
<del>_</del> 3
[5] File Handling

✓ 5.1 Text File

 1 with open("test.txt","w") as file:
       file.write("Hello World")
 1 with open("test.txt","r") as file:
       print(file.read())
→ Hello World
∨ 5.2 CSV File
 1 import csv
 3 with open("test.csv","w",newline='') as file:
            writer = csv.writer(file)
           writer.writerow(["Name","Surname"])
writer.writerow(["Alice","Johnson"])
 5
 6
           writer.writerow(["Bob","Smith"])
 1 import csv
 3 with open("test.csv","r") as file:
       reader = csv.reader(file)
       for row in reader:
 5
          print(row)
    ['Name', 'Surname']
['Alice', 'Johnson']
['Bob', 'Smith']
```

[4] Libraries

```
4.1 Numpy
import numpy library
 1 import numpy as np
ndarray initialization
Construct using python list
 1 # 1d ndarray from 1d python list
 2 list_a1=[1,2,3.5]
 3 arr_a1=np.array(list_a1)
 4 arr_a1
\Rightarrow array([1., 2., 3.5])
 1 # 2d ndarray from 2d python list (list of list)
 2 list_a2=[[1,2],[3,4],[5,6]]
 3 arr_a2=np.array(list_a2)
 4 arr_a2
\rightarrow array([[1, 2],
           [3, 4],
           [5, 6]])
 1 list_a3=[[[1,2],[2,3]],[[3,4],[4,5]]]
 2 arr_a3=np.array(list_a3)
 3 arr a3
[[3, 4],
[4, 5]]])
or construct using some numpy classes and functions
 1 np.zeros(5)
\Rightarrow array([0., 0., 0., 0., 0.])
 1 np.ones((3,4),dtype=float)
1 np.full((4,),999)
⇒ array([999, 999, 999, 999])
 1 np.arange(3,10,2)
\rightarrow array([3, 5, 7, 9])
 1 np.linspace(10,15,11)
→ array([10. , 10.5, 11. , 11.5, 12. , 12.5, 13. , 13.5, 14. , 14.5, 15. ])
 1 np.random.choice(['a','b'],9)
→ array(['a', 'a', 'a', 'b', 'a', 'b', 'a', 'b'], dtype='<U1')</pre>
 1 np.random.randn(10)
array([-0.82222123, 0.7728091, 0.07231017, 0.58226338, 0.25735609, 0.96683725, 1.56468755, -0.36298275, -0.69294433, -1.30699885])
```

1 list_a=[[1,2,3,4],[5,6,7,8],[9,10,11,12]] 2 arr_a=np.array(list_a) 3 arr_a array([[1, 2, 3, 4], [5, 6, 7, 8], [9, 10, 11, 12]]) 1 arr_a.ndim → 2 1 arr_a.shape **→** (3, 4) 1 arr_a.dtype → dtype('int64') 1 arr_a.size → 12 Reshaping & Modification from this original ndarray 1 arr_a array([[1, 2, 3, 4], [5, 6, 7, 8], [9, 10, 11, 12]]) try to convert into 3D array 1 arr_a.reshape((2,2,3)) ⇒ 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 1 arr_a.reshape((-1,6)) \rightarrow array([[1, 2, 3, 4, 5, 6], [7, 8, 9, 10, 11, 12]]) Would you like to try this? 1 arr_a.reshape((-1,5)) ValueError Traceback (most recent call last) <ipython-input-127-286d5aa6424c> in <cell line: 0>() ---> 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.

ndarray properties

Ans: To reshape the array into a shape with 5 columns, the total number of elements must be divisible by 5. However, 12 is not divisible by 5.

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

```
1 np.append(arr_a,13)
\Rightarrow array([ 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13])
 1 np.append(arr_a,arr_a[0])
\Rightarrow array([ 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 1, 2, 3, 4])
 1 np.append(arr_a,arr_a[0].reshape((1,-1)),axis=0)
1 np.append(arr_a,arr_a[:,0].reshape((-1,1)),axis=1)
array([[ 1, 2, 3, 4, 1], [ 5, 6, 7, 8, 5], [ 9, 10, 11, 12, 9]])
 1 np.concatenate([arr_a,arr_a])
\rightarrow 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]])
 1 np.concatenate([arr_a,arr_a],axis=1)
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
 1 arr_a
⇒ array([[ 1, 2, 3, 4], [ 5, 6, 7, 8], [ 9, 10, 11, 12]])
try to access all element at the first row
 1 arr_a[1]
\rightarrow array([5, 6, 7, 8])
then you would like to access the second element from the first row
 1 arr_a[1][2]
1 arr_a[1,2]
<del>_____</del> 7
Next, try to access all element start from 1th in the first row
```

⇒ array([6, 7, 8])

1 arr_a[1,1:]

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

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

sometimes you may specify some row number using list within indicing

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

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

∨ Boolean slicing

based on this original array

try to filter all elements which more than 5

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

Run the cell below and answer a question.

```
1 arr_a[(arr_a>5)&(arr_a<10)]

array([6, 7, 8, 9])
```

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

Ans: The output consists of only the elements of arr_a that correspond to True in the combined boolean mask (arr a > 5) & (arr a < 10).

Try running the cell below.

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

Ans: The and operator in Python is a logical operator that works with single boolean values. While arr_a > 5 and arr_a < 10 are both boolean arrays, not single boolean values.

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

Ans: To perform element-wise logical AND on NumPy arrays, try using the & operator instead of and >> arr_a[(arr_a>5) & (arr_a<10)]

This is some operations for only 1 array

```
1 np.sqrt(arr_b)
```

This is some operations for 2 arrays with the same shape

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

```
1 arr_a*3
```

1 1+arr_a**2

Try to play with 2 arrays with different shape

→ Basic aggregations

```
1 arr_a
```

```
1 arr_a.sum()
<del>→</del> 78
 1 arr_a.mean()
→ 6.5
 1 arr_a.min()
<u>→</u> 1
 1 arr_a.max()
<del>→</del> 12
 1 arr_a.std() #standard division
3.452052529534663

✓ ndarray axis

 1 arr_a
1 arr_a.sum(axis=0)
⇒ array([15, 18, 21, 24])
 1 arr_a.sum(axis=1)
→ 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: axis=0: is performed column-wise. and axis=1: is performed row-wise.
4.2 Pandas

✓ Series

 1 import pandas as pd
 2 import numpy as np
 1 pd.Series(np.random.randn(6))
\overline{z}
     0 0.637273
     1 1.327869
     2 -0.620961
     3 -0.203078
     4 0.378510
     5 0.036953
    dtyne: float64
 1 pd.Series(np.random.randn(6), index=['a','b','c','d','e','f'])
```

Constructing Dataframe

Constructing DataFrame from a dictionary

```
1 d = {'col1':[1,2], 'col2': [3,4]}
1 df = pd.DataFrame(data=d)
2 df
     col1 col2
       2 4
1 d2 = {'Name':['Joe','Nat','Harry','Sam','Monica'],
        'Age': [20,21,19,20,22]}
1 df2 = pd.DataFrame(data=d2)
2 df2
\overline{\Rightarrow}
         Name Age
     0
          Joe 20
     1
          Nat 21
     2
        Harry 19
     3
               20
         Sam
               22
     4 Monica
```

Constructing DataFrame from a List

Creating DataFrame from file

```
1 # Read csv file from path and store to df for create dataframe
2 df = pd.read_csv('nss15.csv')
```



	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bodyPart	disposition	location	product
0	150733174	7/11/2015	15.7762	V	5.0	Male	NaN	57.0	33.0	1.0	9.0	1267.0
1	150734723	7/6/2015	83.2157	S	36.0	Male	White	57.0	34.0	1.0	1.0	1439.0
2	150817487	8/2/2015	74.8813	L	20.0	Female	NaN	71.0	94.0	1.0	0.0	3274.0
3	150717776	6/26/2015	15.7762	V	61.0	Male	NaN	71.0	35.0	1.0	0.0	611.0
4	150721694	7/4/2015	74.8813	L	88.0	Female	Other	62.0	75.0	1.0	0.0	1893.0
234085	150968928	9/22/2015	15.7762	V	23.0	Male	Black	64.0	92.0	1.0	1.0	1141.0
234086	150965850	9/24/2015	83.2157	S	37.0	Female	NaN	64.0	31.0	1.0	1.0	4014.0
234087	150971407	9/26/2015	5.6748	С	13.0	Male	White	71.0	79.0	1.0	9.0	1211.0
234088	151026924	10/6/2015	5.6748	С	1.0	Male	White	53.0	75.0	1.0	1.0	4057.0
234089	15100638	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
234090 ro	ws × 12 colum	ins										

∨ Viewing DataFrame information

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

Check simple information

- 1 # Check dimension by .shape
- 2 df.shape
- → (234090, 12)
- 1 # Display the first 5 rows by default
 2 df.head()

₹	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bodyPart	disposition	location	product
(150733174	7/11/2015	15.7762	V	5.0	Male	NaN	57.0	33.0	1.0	9.0	1267.0
1	150734723	7/6/2015	83.2157	S	36.0	Male	White	57.0	34.0	1.0	1.0	1439.0
2	150817487	8/2/2015	74.8813	L	20.0	Female	NaN	71.0	94.0	1.0	0.0	3274.0
3	150717776	6/26/2015	15.7762	V	61.0	Male	NaN	71.0	35.0	1.0	0.0	611.0
4	150721694	7/4/2015	74.8813	L	88.0	Female	Other	62.0	75.0	1.0	0.0	1893.0
4												

- 1 # Display the first 3 rows
- 2 df.head(3)

-		caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bodyPart	disposition	location	product
(0	150733174	7/11/2015	15.7762	V	5.0	Male	NaN	57.0	33.0	1.0	9.0	1267.0
,	1	150734723	7/6/2015	83.2157	S	36.0	Male	White	57.0	34.0	1.0	1.0	1439.0
1	2	150817487	8/2/2015	74.8813	L	20.0	Female	NaN	71.0	94.0	1.0	0.0	3274.0
4													

1 # Display the last 5 rows by default
2 df.tail()

₹		caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bodyPart	disposition	location	product	
	234085	150968928	9/22/2015	15.7762	V	23.0	Male	Black	64.0	92.0	1.0	1.0	1141.0	
	234086	150965850	9/24/2015	83.2157	S	37.0	Female	NaN	64.0	31.0	1.0	1.0	4014.0	
	234087	150971407	9/26/2015	5.6748	С	13.0	Male	White	71.0	79.0	1.0	9.0	1211.0	
	234088	151026924	10/6/2015	5.6748	С	1.0	Male	White	53.0	75.0	1.0	1.0	4057.0	
	234089	15100638	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	4													>

- $\ensuremath{\text{1}}\xspace\#\ensuremath{\text{4}}\xspace$ Overview information of dataframe
- 2 df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 234090 entries, 0 to 234089 Data columns (total 12 columns): Non-Null Count Dtype # Column 234090 non-null int64 0 caseNumber treatmentDate 234089 non-null object statWeight 234089 non-null float64 stratum 234089 non-null object 4 age 234089 non-null float64 sex 234087 non-null object race 143318 non-null object diagnosis 234089 non-null float64 bodyPart 234089 non-null float64 disposition 234089 non-null float64 location 234089 non-null float64 10 location 11 product 234089 non-null float64 dtypes: float64(7), int64(1), object(4) memory usage: 21.4+ MB Select column, multiple column, with condition 1 df.columns Index(['caseNumber', 'treatmentDate', 'statWeight', 'stratum', 'age', 'sex', 'race', 'diagnosis', 'bodyPart', 'disposition', 'location', 'product'], dtype='object') 1 #select single column 2 df['age'] $\overline{\mathcal{D}}$ age 0 5.0 36.0 1 2 20.0 3 61.0

age
0 5.0
1 36.0
2 20.0
3 61.0
4 88.0
...
234085 23.0
234086 37.0
234087 13.0
234088 1.0
234089 NaN
234090 rows × 1 columns

dtype: float64

1 df.age

 $\overline{\Rightarrow}$ age 0 5.0 36.0 1 2 20.0 3 61.0 88.0 4 ... **234085** 23.0 **234086** 37.0 **234087** 13.0 234088 1.0 234089 NaN

234090 rows × 1 columns

dtyne: float64

1 #select multiple column

2 df[['treatmentDate','statWeight','age','sex']]

	treatmentDate	statWeight	age	sex
0	7/11/2015	15.7762	5.0	Male
1	7/6/2015	83.2157	36.0	Male
2	8/2/2015	74.8813	20.0	Female
3	6/26/2015	15.7762	61.0	Male
4	7/4/2015	74.8813	88.0	Female
234085	9/22/2015	15.7762	23.0	Male
234086	9/24/2015	83.2157	37.0	Female
234087	9/26/2015	5.6748	13.0	Male
234088	10/6/2015	5.6748	1.0	Male
234089	NaN	NaN	NaN	NaN
234090 rc	ws × 4 columns			

Viewing the unique value

```
1 df.race.unique()
```

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

Describe

```
1 df['age'].describe()
```

_		
→		age
	count	234089.000000
	mean	31.323274
	std	26.077750
	min	0.000000
	25%	10.000000
	50%	23.000000
	75%	51.000000
	max	106.000000

Select row with condition

```
1 #select by condition
```

² df[df['sex'] == 'Male']

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bodyPart	disposition	location	product
0	150733174	7/11/2015	15.7762	V	5.0	Male	NaN	57.0	33.0	1.0	9.0	1267.0
1	150734723	7/6/2015	83.2157	S	36.0	Male	White	57.0	34.0	1.0	1.0	1439.0
3	150717776	6/26/2015	15.7762	V	61.0	Male	NaN	71.0	35.0	1.0	0.0	611.0
6	150713483	6/8/2015	15.7762	V	25.0	Male	Black	51.0	33.0	4.0	9.0	1138.0
7	150704114	6/14/2015	83.2157	S	53.0	Male	White	57.0	30.0	1.0	0.0	5040.0
234081	150953450	9/19/2015	5.6748	С	7.0	Male	White	57.0	80.0	4.0	0.0	3286.0
234084	150906288	8/27/2015	5.6748	С	14.0	Male	White	57.0	33.0	1.0	9.0	1329.0
234085	150968928	9/22/2015	15.7762	V	23.0	Male	Black	64.0	92.0	1.0	1.0	1141.0
234087	150971407	9/26/2015	5.6748	С	13.0	Male	White	71.0	79.0	1.0	9.0	1211.0
234088	151026924	10/6/2015	5.6748	С	1.0	Male	White	53.0	75.0	1.0	1.0	4057.0
127729 rd	ows × 12 colum	nns										
,												

1 #select by multiple condition

2 df[(df['sex'] == 'Male') & (df['age'] > 80)]

	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bodyPart	disposition	location	product
8	150736558	7/16/2015	83.2157	S	98.0	Male	Black	59.0	76.0	1.0	1.0	1807.0
63	150418623	1/12/2015	15.0591	V	97.0	Male	Other	62.0	75.0	4.0	1.0	4076.0
97	150700375	6/28/2015	83.2157	S	85.0	Male	NaN	59.0	92.0	1.0	0.0	478.0
131	150940801	9/14/2015	15.7762	V	96.0	Male	NaN	62.0	75.0	1.0	5.0	1807.0
177	160110774	12/19/2015	85.7374	S	81.0	Male	White	59.0	82.0	1.0	1.0	3278.0
233609	150117332	1/2/2015	78.5926	S	82.0	Male	NaN	59.0	76.0	4.0	9.0	3223.0
233867	150822235	7/25/2015	15.7762	V	93.0	Male	NaN	57.0	79.0	4.0	5.0	1679.0
233932	151029869	6/17/2015	74.8813	L	85.0	Male	NaN	71.0	36.0	4.0	1.0	4076.0
233985	151015969	10/6/2015	83.2157	S	82.0	Male	NaN	62.0	75.0	1.0	5.0	1807.0
234052	150827615	7/30/2015	15.7762	V	85.0	Male	NaN	59.0	33.0	1.0	1.0	1884.0
4407 rows	x 12 columns											
1												

Select row with .iloc

1 # select row by .iloc
2 df.iloc[10:15]

₹	caseNumber	treatmentDate	statWeight	stratum	age	sex	race	diagnosis	bodyPart	disposition	location	product
10	150734952	7/4/2015	15.7762	V	20.0	Male	Black	59.0	82.0	1.0	1.0	1894.0
11	150821622	7/20/2015	83.2157	S	20.0	Female	White	57.0	36.0	1.0	9.0	1267.0
12	150713631	7/4/2015	15.7762	V	11.0	Male	NaN	60.0	88.0	1.0	0.0	3274.0
13	150666343	6/27/2015	15.7762	V	26.0	Female	White	62.0	75.0	1.0	1.0	1807.0
14	150748843	7/16/2015	37.6645	L	33.0	Male	Asian	53.0	93.0	1.0	1.0	4057.0
4												

1 # select column by .iloc

2 df.iloc[:,[0,1,2,3,4]]

\Rightarrow		caseNumber	treatmentDate	statWeight	stratum	age
	0	150733174	7/11/2015	15.7762	V	5.0
	1	150734723	7/6/2015	83.2157	S	36.0
	2	150817487	8/2/2015	74.8813	L	20.0
	3	150717776	6/26/2015	15.7762	V	61.0
	4	150721694	7/4/2015	74.8813	L	88.0
Selec	t columr	n and row witl	h .loc			
	234085	150968928	9/22/2015	15.7762	V	23.0
1 #	select	column and	l low by .loc			

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

→		treatmentDate	statWeight	stratum	age	sex	race	diagnosis
	0	7/11/2015	15.7762	V	5.0	Male	NaN	57.0
	1	7/6/2015	83.2157	S	36.0	Male	White	57.0
	2	8/2/2015	74.8813	L	20.0	Female	NaN	71.0
	3	6/26/2015	15.7762	V	61.0	Male	NaN	71.0
	4	7/4/2015	74.8813	L	88.0	Female	Other	62.0
	5	7/2/2015	5.6748	С	1.0	Female	White	71.0
	6	6/8/2015	15.7762	V	25.0	Male	Black	51.0

1 # select row by condition

2
3 df.loc[df['age']>80, ['treatmentDate', 'age']]

⇒÷		treatmentDate	age
	4	7/4/2015	88.0
	8	7/16/2015	98.0
	39	5/3/2015	88.0
	46	4/15/2015	91.0
	63	1/12/2015	97.0
	233985	10/6/2015	82.0
	234047	8/4/2015	83.0
	234049	6/16/2015	85.0
	234052	7/30/2015	85.0
	234069	8/29/2015	83.0

14171 rows × 2 columns

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

Ans: .iloc is Integer-based indexing (position-based). While .loc is Label-based indexing (name-based).

