

1. Differences between `loc` and `iloc`

The main distinction between `loc` and `iloc` is:

- `loc` is label-based, which means that you have to specify rows and columns based on their row and column **labels**.
- `iloc` is integer position-based, so you have to specify rows and columns by their **integer position values** (0-based integer position).

Here are some differences and similarities between `loc` and `iloc` :

	loc	iloc
A value	A single label or integer e.g. <code>loc[A]</code> or <code>loc[1]</code>	A single integer e.g. <code>iloc[1]</code>
A list	A list of labels e.g. <code>loc[[A, B]]</code>	A list of integers e.g. <code>iloc[[1, 2, 3]]</code>
Slicing	e.g. <code>loc[A:B]</code> , A and B are included	e.g. <code>iloc[n:m]</code> , n is included, m is excluded
Conditions	A bool Series or list	A bool list
Callable function	<code>loc[lambda x: x[2]]</code>	<code>iloc[lambda x: x[2]]</code>

Differences and Similarities between loc and iloc (image by author)

For demonstration, we create a DataFrame and load it with the **Day** column as the index.

```
df = pd.read_csv('data/data.csv', index_col=['Day'])
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

image by author

2. Selecting via a single value

Both `loc` and `iloc` allow input to be a single value. We can use the following syntax for data selection:

- `loc[row_label, column_label]`
- `iloc[row_position, column_position]`

For example, let's say we would like to retrieve Friday's temperature value.

With `loc`, we can pass the row label `'Fri'` and the column label `'Temperature'`.

```
# To get Friday's temperature
>>> df.loc['Fri', 'Temperature']

10.51
```

The equivalent `iloc` statement should take the row number `4` and the column number `1`.

```
# The equivalent `iloc` statement
>>> df.iloc[4, 1]

10.51
```

We can also use `:` to return all data. For example, to get all rows:

```
# To get all rows
>>> df.loc[:, 'Temperature']

Day
Mon    12.79
Tue    19.67
Wed    17.51
Thu    14.44
Fri     9.51
Sat    11.07
Sun    17.50
Name: Temperature, dtype: float64

# The equivalent `iloc` statement
>>> df.iloc[:, 1]
```

And to get all columns:

```
# To get all columns
>>> df.loc['Fri', :]

Weather      Shower
Temperature   10.51
Wind          26
Humidity      79
Name: Fri, dtype: object

# The equivalent `iloc` statement
>>> df.iloc[4, :]
```

Note that the above 2 outputs are **Series**. `loc` and `iloc` will return a **Series** when the result is 1-dimensional data.

3. Selecting via a list of values

We can pass a list of labels to `loc` to select multiple rows or columns:

```
# Multiple rows
>>> df.loc[['Thu', 'Fri'], 'Temperature']

Day
Thu    14.44
Fri    10.51
Name: Temperature, dtype: float64
```

```
# Multiple columns
>>> df.loc['Fri', ['Temperature', 'Wind']]

Temperature    10.51
Wind           26
Name: Fri, dtype: object
```

Similarly, a list of integer values can be passed to `iloc` to select multiple rows or columns. Here are the equivalent statements using `iloc`:

```
>>> df.iloc[[3, 4], 1]

Day
Thu    14.44
Fri    10.51
Name: Temperature, dtype: float64

>>> df.iloc[4, [1, 2]]

Temperature    10.51
Wind           26
Name: Fri, dtype: object
```

All the above outputs are **Series** because their results are 1-dimensional data.

The output will be a **DataFrame** when the result is 2-dimensional data, for

example, to access multiple rows and columns

```
# Multiple rows and columns  
rows = ['Thu', 'Fri']  
cols=['Temperature','Wind']  
  
df.loc[rows, cols]
```

	Temperature	Wind
Day		
Thu	14.44	11
Fri	10.51	26

The equivalent `iloc` statement is:

```
rows = [3, 4]  
cols = [1, 2]  
  
df.iloc[rows, cols]
```


4. Selecting a range of data via slice

Slice (written as `start:stop:step`) is a powerful technique that allows selecting a range of data. It is very useful when we want to select everything in between two items.

`loc` **with slice**

With `loc`, we can use the syntax `A:B` to select data from label A to label B (Both A and B are included):

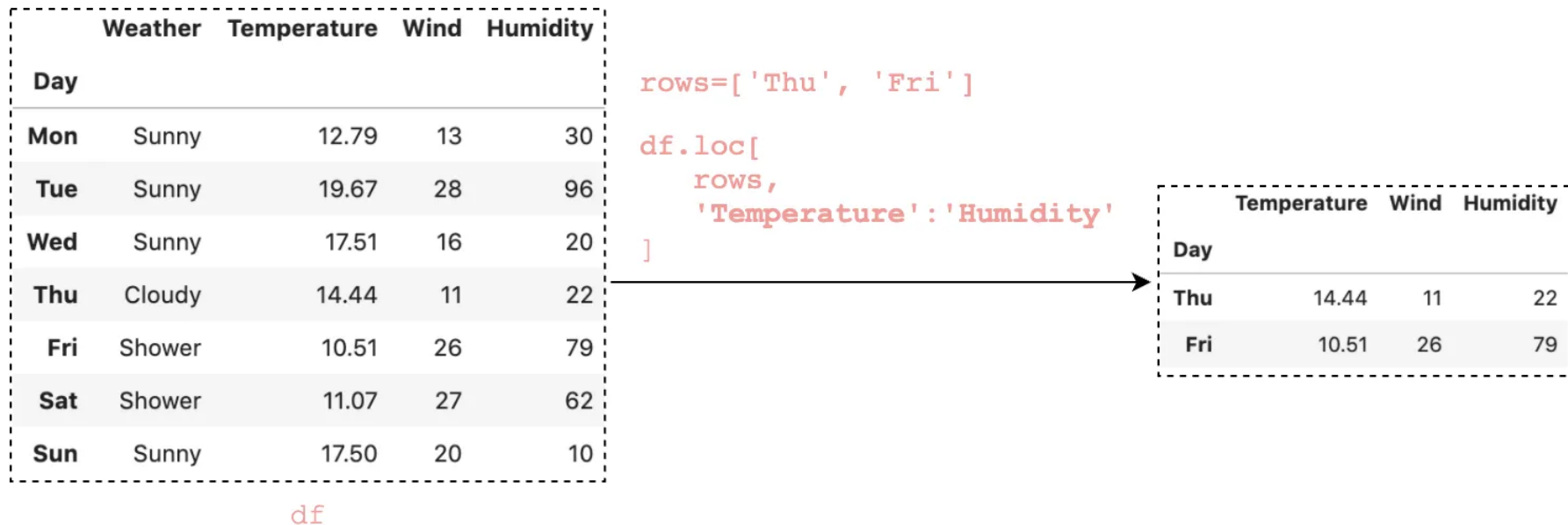


image by author

```
# Slicing row labels  
cols = ['Temperature', 'Wind']  
  
df.loc['Mon':'Thu', cols]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

```
cols = ['Temperature', 'Wind']  
df.loc['Mon':'Thu', cols]
```

	Temperature	Wind
Day		
Mon	12.79	13
Tue	19.67	28
Wed	17.51	16
Thu	14.44	11

image by author

We can use the syntax `A:B:S` to select data from label A to label B with step size S (Both A and B are included):

```
# Slicing with step  
df.loc['Mon':'Fri':2, :]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

df.loc['Mon':'Fri':2, :]

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Wed	Sunny	17.51	16	20
Fri	Shower	10.51	26	79

image by author

iloc with slice

With `iloc`, we can also use the syntax `n:m` to select data from position **n** (included) to position **m** (excluded). However, the main difference here is that the endpoint (**m**) is excluded from the `iloc` result.

For example, selecting columns from position 0 up to 3 (excluded):

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

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

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

	Weather	Temperature	Wind
Day			
Tue	Sunny	19.67	28
Wed	Sunny	17.51	16

image by author

Similarly, we can use the syntax `n:m:s` to select data from position **n** (included) to position **m** (excluded) with step size **s**. Notes that the endpoint **m** is excluded.

```
df.iloc[0:4:2, :]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

```
df.iloc[0:4:2, :]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Wed	Sunny	17.51	16	20

image by author

5. Selecting via conditions and callable

Conditions

`loc` with conditions

Often we would like to filter the data based on conditions. For example, we

may need to find the rows where humidity is greater than 50.

With `loc`, we just need to pass the condition to the `loc` statement.

```
# One condition  
df.loc[df.Humidity > 50, :]
```

The diagram illustrates the use of `df.loc` to filter data based on a single condition. It shows a full DataFrame on the left and a filtered DataFrame on the right, connected by an arrow. The condition `df.Humidity > 50` is written in red text above the arrow.

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

`df`

`df.loc[df.Humidity > 50, :]`

	Weather	Temperature	Wind	Humidity
Day				
Tue	Sunny	19.67	28	96
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62

image by author

Sometimes, we may need to use multiple conditions to filter our data. For example, find all the rows where humidity is more than 50 and the weather

is Shower:

```
## multiple conditions
df.loc[
    (df.Humidity > 50) & (df.Weather == 'Shower'),
    ['Temperature', 'Wind'],
]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

```
df.loc[
    (df.Humidity > 50) & (df.Weather == 'Shower'),
    ['Temperature', 'Wind'],
]
```

	Temperature	Wind
Day		
Fri	10.51	26
Sat	11.07	27

image by author

iloc with conditions

For `iloc`, we will get a `ValueError` if pass the condition straight into the

statement:

```
# Getting ValueError  
df.iloc[df.Humidity > 50, :]
```

```
-----  
ValueError                                Traceback (most recent call last)  
~/anaconda3/envs/tf-tutorial/lib/python3.7/site-packages/pandas/core/indexing.py  
    701         try:  
--> 702             self._validate_key(k, i)  
    703         except ValueError as err:  
  
~/anaconda3/envs/tf-tutorial/lib/python3.7/site-packages/pandas/core/indexing.py  
    1343         raise ValueError(  
-> 1344             "iLocation based boolean indexing cannot use "  
    1345             "an indexable as a mask"  
  
ValueError: iLocation based boolean indexing cannot use an indexable as a mask
```

image by author

We get the error because `iloc` cannot accept a boolean Series. It only accepts a boolean list. We can use the `list()` function to convert a Series into a boolean list.


```
# Single condition
df.iloc[list(df.Humidity > 50)]
```

Similarly, we can use `list()` to convert the output of multiple conditions into a boolean list:

```
## multiple conditions
df.iloc[
    list((df.Humidity > 50) & (df.Weather == 'Shower')),
    :,
]
```

Callable function

`loc` with callable

`loc` accepts a **callable** as an indexer. The callable must be a function with one argument that returns valid output for indexing.

For example to select columns

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

```
df.loc[
    :,
    lambda df: ['Humidity', 'Wind']
]
```

	Humidity	Wind
Day		
Mon	30	13
Tue	96	28
Wed	20	16
Thu	22	11
Fri	79	26
Sat	62	27
Sun	10	20

And to filter data with a callable:

```
# With condition
df.loc[lambda df: df.Humidity > 50, :]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

```
df.loc[
    lambda df: df.Humidity > 50,
    :
]
```

	Weather	Temperature	Wind	Humidity
Day				
Tue	Sunny	19.67	28	96
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62

image by author

iloc with callable

iloc can also take a **callable** as an indexer.

```
df.iloc[lambda df: [0,1], :]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

df.iloc[lambda df: [0,1], :]

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96

image by author

To filter data with callable, `iloc` will require `list()` to convert the output of conditions into a boolean list:

```
df.iloc[lambda df: list(df.Humidity > 50), :]
```

	Weather	Temperature	Wind	Humidity
Day				
Mon	Sunny	12.79	13	30
Tue	Sunny	19.67	28	96
Wed	Sunny	17.51	16	20
Thu	Cloudy	14.44	11	22
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62
Sun	Sunny	17.50	20	10

df

```
df.iloc[
    lambda df: list(df.Humidity > 50),
    :
]
```

	Weather	Temperature	Wind	Humidity
Day				
Tue	Sunny	19.67	28	96
Fri	Shower	10.51	26	79
Sat	Shower	11.07	27	62

image by author

6. loc and iloc are interchangeable when labels are 0-based integers

For demonstration, let's create a DataFrame with 0-based integers as headers and index labels.

```
df = pd.read_csv(  
    'data/data.csv',  
    header=None,  
    skiprows=[0],  
)
```

With `header=None`, the Pandas will generate 0-based integer values as headers. With `skiprows=[0]`, those headers **Weather, Temperature**, etc we have been using will be skipped.

	0	1	2	3	4
0	Mon	Sunny	12.79	13	30
1	Tue	Sunny	19.67	28	96
2	Wed	Sunny	17.51	16	20
3	Thu	Cloudy	14.44	11	22
4	Fri	Shower	10.51	26	79
5	Sat	Shower	11.07	27	62
6	Sun	Sunny	17.50	20	10

image by author

Now, `loc`, a label-based data selector, can accept a single integer and a list of integer values. For example:

```
>>> df.loc[1, 2]  
19.67
```

```
>>> df.loc[1, [1, 2]]  
1    Sunny  
2    19.67  
Name: 1, dtype: object
```

The reason they are working is that those integer values (`1` and `2`) are interpreted as *labels* of the index. This use is **not** an integer position along with the index and is a bit confusing.

In this case, `loc` and `iloc` are interchangeable when selecting via a single value or a list of values.

```
>>> df.loc[1, 2] == df.iloc[1, 2]  
True
```



```
>>> df.loc[1, [1, 2]] == df.iloc[1, [1, 2]]
1      True
2      True
Name: 1, dtype: bool
```

Note that `loc` and `iloc` will return different results when selecting via slice and conditions. They are essentially different because:

- slice: endpoint is excluded from `iloc` result, but included in `loc`
- conditions: `loc` accepts boolean Series, but `iloc` can only accept a boolean list.

Conclusion

Finally, here is a summary

`loc` is label based and allowed inputs are:

- A single label 'A' or 2 (Note that 2 is interpreted as a *label* of the index.)
- A list of labels ['A', 'B', 'C'] or [1, 2, 3] (Note that 1, 2, 3 are interpreted as *labels* of the index.)

- A slice with labels 'A':'C' (Both are included)
- Conditions, a boolean Series or a boolean array
- A `callable` function with one argument

`iloc` is integer position based and allowed inputs are:

- An integer e.g. `2` .
- A list or array of integers `[1, 2, 3]` .
- A slice with integers `1:7` (the endpoint `7` is excluded)
- Conditions, but only accept a boolean array
- A `callable` function with one argument

`loc` and `iloc` are interchangeable when the labels of Pandas DataFrame are 0-based integers

I hope this article will help you to save time in learning Pandas data selection. I recommend you to check out the [documentation](#) to know about other things you can do.