Lecture 9 Introduction to Pandas

<u>Pandas--Python Data Analysis Library (https://pandas.pydata.org/)</u> provides the high-performance, easy-to-use data structures and data analysis tools in Python, which is very useful in Data Science. In our lectures, we only focust on the <u>elementary usages (https://pandas.pydata.org/pandas-docs/stable/user_guide/10min.html)</u>.

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

In [ ]: pd.__version__
In [ ]: dir(pd)
```

Important Concepts: Series and DataFrame

In short, Series represents one variable (attributes) of the datasets, while DataFrame represents the whole tabular data (it also supports multi-index or tensor cases -- we will not discuss these cases here).

Series is Numpy 1d array-like, additionally featuring for "index" which denotes the sample name, which is also similar to Python built-in dictionary type.

```
In [ ]: s1 = pd.Series([2, 4, 6])
In [ ]: type(s1)
In [ ]: s1.index
In [ ]: s2 = pd.Series([2, 4, 6],index = ['a','b','c'])
In [ ]: s2
In [ ]: s2
In [ ]: s2_num = s2.values # change to Numpy -- can be view instead of copy if the elements a re all numbers s2_num
In [ ]: np.shares_memory(s2_num,s2)
In [ ]: s2_num_copy = s2.to_numpy(copy = True) # more recommended in new version of Pandas -- can specify view/copy np.shares_memory(s2_num_copy,s2)
```

Selection by position

```
In [ ]: s2[0:2]
```

Selection by index (label)

```
In [ ]: s2['a']
s2[['a','b']]
```

Create the pandas DataFrame from Series . Note that in Pandas, the row/column of DataFrame are termed as index and columns .

Creating DataFrame from Files

```
In [ ]: house_price = pd.read_csv('kc_house_data.csv')
house_price
In [ ]: house_price.shape # dimension of the data
In [ ]: house_price.info() # basic dataset information
In [ ]: house_price.head(3) # show the head lines
In [ ]: house_price.sample(5) # show the random samples
In [ ]: house_price.describe() # descriptive statistics
```

```
In [ ]: head = house_price.head()
   head.to_csv('head.csv')

In [ ]: head.sort_values(by='price')

In [ ]: help(head.sort_values)

In [ ]: head.to_numpy()

In [ ]: help(head.to_numpy)
```

Selection

Selection by label (.loc) or by position (.iloc)

First recall the basic slicing for Series

```
In [ ]: s2
In [ ]: s2[0:2] # by position
In [ ]: s2['a':'c'] # by label
In [ ]: s2.index
```

However, confusions may occur if the "labels" are very similar to "position"

```
In [ ]: s3= pd.Series(['a','b','c','d','e'])
s3
In [ ]: s3.index
In [ ]: s3[0:2] #slicing -- this is confusing, although it is still by position
```

That's why pandas use .loc and .iloc to strictly distinguish by label or by position.

```
In [ ]: s3.loc[0:2] # by label
In [ ]: s3.iloc[0:2] # by position
```

The same applies to DataFrame.

```
In [ ]: head
In [ ]: head.iloc[:3,:2]
In [ ]: head.loc[:3,:'date']
```

```
In [ ]: help(head.loc)
In [ ]: help(head.iloc)
```

More Comments on Slicing and Indexing in DataFrame

Slicing picks rows, while indexing picks columns -- this can be confusing, and that's why .iloc and .loc are more strict.

```
In [ ]: head['date'] #same with head.date
In [ ]:
       head[['date','price']]
In [ ]:
       head[['date']]
In [ ]: head[0:2] #slicing
In [ ]:
       | head['date':'price'] # this is wrong
In [ ]: head[:,'date':'price']# this is also wrong!
       head[:,['date','price']] # this is also wrong!!
In []: head[1:3][['date','price']] # to do slicing and indexing "simultaneously", you have t
        o do them separately!
In [ ]: head.loc[:,'date':'price'] # no problem for slicing in .loc
       | head.loc[:,['date','price']] # fancy indexing is also supported in .loc
In [ ]:
        states
In [ ]: states['California':'Texas']
In [ ]:
        states['population']
In [ ]: states['California':'Texas', 'population'] # this is wrong
In [ ]: states.loc['California':'Texas','population']
       states.loc['California':'Texas']
In [ ]:
```

Boolean Selection

```
In [ ]: ind = states.area>200000
ind

In [ ]: states[ind]

In [ ]: states[ind, 'area'] # this is wrong!

In [ ]: states[ind]['area']

In [ ]: states.loc[states.area>200000, 'population'] # equivalently, states.loc[ind, 'population']
```

```
In [ ]: states.iloc[ind.to_numpy(),1] # in iloc, the boolen should be the Numpy array
In [ ]: random
In [ ]: random[random['foo']>0.6]
In [ ]: house_price
```

Sometimes it's very useful to use the isin method to filter samples.