# Lecture 9 Introduction to Pandas

Pandas--*Python Data Analysis Library* (https://pandas.pydata.org/) 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 elementary usages (https://pandas.pydata.org/pandas-docs/stable/user_guide/10min.html).

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_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']]
```

`Series` and Python Dictionary

In [ ]:
```
population_dict = {'California': 38332521,
                   'Texas': 26448193,
                   'New York': 19651127,
                   'Florida': 19552860,
                   'Illinois': 12882135} # this is the built-in python dictionary
population = pd.Series(population_dict) # initialize Series with dictionary
population
```

In [ ]:
```
population_dict['Texas'] # key and value
```

In [ ]:
```
area_dict = {'California': 423967, 'Texas': 695662, 'New York': 141297,
             'Florida': 170312, 'Illinois': 149995}
area = pd.Series(area_dict)
area
```

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

In [ ]:
```
states = pd.DataFrame({'population': population,
                       'area': area}) # variable names
states
```

In [ ]:
```
type(states)
```

In [ ]:
```
states.index
```

In [ ]:
```
states.columns
```

In [ ]:
```
states['area']
```

In [ ]:
```
states.area
```

In [ ]:
```
type(states['area'])
```

In [ ]:
```
random = pd.DataFrame(np.random.rand(3, 2),columns=['foo', 'bar'],index=['a', 'b',
'c'])
random
```

In [ ]:
```
random.T
```

## 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
```

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']
```

*Note: in the latest version of Pandas, the mixing selection .ix is **deprecated** -- note this when reading the Data Science Handbook!*

## 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[0:2] #slicing
```

In [ ]:
```
head['date':'price'] # this is wrong
```

In [ ]:
```
head[:,'date':'price']# this is also wrong!
```

In [ ]:
```
head[:,['date','price']] # this is also wrong!!
```

In [ ]:
```
head.loc[:,'date':'price']
```

In [ ]:
```
states
```

In [ ]:
```
states['California':'Texas']
```

In [ ]:
```
states['population']
```

In [ ]:
```
states['California':'Texas','population'] # this is wrong
```

In [ ]:
```
states.loc['California':'Texas','population']
```

In [ ]:
```
states.loc['California':'Texas']
```

## Boolean Selection

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

In [ ]:
```
states[ind]
```

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

In [ ]:
```
random
```

In [ ]:
```
random[random['foo']>0.6]
```

In [ ]:
```
house_price
```

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

In [ ]:
```
house_price[house_price.loc[:,'bedrooms'].isin([2,4])]
```

In [ ]:
```
house_price[house_price['bedrooms'].isin([2,4])] # the same with column index
```