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MIE1624H – Introduction to Data Science and Analytics

Lecture 2 – Python Programming

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Lecture outline

Introduction to Data Science and Analytics (continuing Lecture 1)

Python essentials

- IPython notebooks
- Modules
- Variables and types
- Operators and comparisons
- Compound types - strings, tuples, lists and dictionaries
- Control flow - conditional statements (if, elif, else), loops
- Functions
- Classes
- Files and the operating system
- Exception handling

Lecture outline

Introduction to Pandas

- Introduction to pandas data structures – DataFrame, index objects
- Pandas essential functionality
- Summarizing and computing descriptive statistics
- Pivot tables in pandas

Web-scrapping with Python

Python Essentials

Roadmap

Python essentials

Variables and types

Operators and comparisons

Compound types - strings, tuples, lists and dictionaries

Functions

Modules

Files and the operating system

Control flow - conditional statements (if, elif, else), loops

Exception handling

Introduction to Pandas

Introduction to pandas data structures – DataFrame, index objects

Pandas essential functionality

Summarizing and computing descriptive statistics

Pivot tables in pandas

IPython notebooks

Introducing Python

Python is an interpreted language (not a compiled one)

=> you can run code incrementally, one statement at a time

Expressions and Values

Arithmetic Operators

Operator	Operation	Expression	English description	Result
+	addition	11 + 56	11 plus 56	67
-	subtraction	23 - 52	23 minus 52	-29
*	multiplication	4 * 5	4 multiplied by 5	20
**	exponentiation	2 ** 5	2 to the power of 5	32
/	division	9 / 2	9 divided by 2	4.5
//	integer division	9 // 2	9 divided by 2	4
%	modulo (remainder)	9 % 2	9 mod 2	1

Arithmetic Operator Precedence - 1

When multiple operators are combined in a single expression, the operations are evaluated in order of precedence.

Operator	Precedence
**	highest
- (negation)	
*, /, //, %	
+ (addition), - (subtraction)	lowest

Arithmetic Operator Precedence - 2

```
>>> 10 % 3
1
>>> 25 + 30 / 6
30.0
>>> 5 + 4 ** 2
21
>>> 100 - 25 * 3 % 4
97
>>> 3 + 2 + 1 - 5 + 4 % 2 - 1 / 4 + 6
6.75
```

Types

A ***type*** is a set of *values* and the *operations* that can be performed on those values.

Types `int` and `float`

`int`: integer

3, 4, 894, 0, -3, -18

`float`: floating point number (an approximation to a real number)

5.6, 7.342, 53452.0, 0.0, -89.34, -9.5

```
>>> 5 + 2 * 4
13
>>> 5.0 + 2 * 4
13.0
>>> 30/6
5.0
>>> 30//6
5
```

Type str - 1

- A *string literal* is a sequence of characters. A string can be made up of letters, numbers, and special characters.
- Strings in Python start and end with a single quote (') or double quotes (").
- If a string begins with a single quote, it must end with a single quote. The same applies to double-quoted strings.
- **The two types of quotes cannot be mixed!**

Type str - 2

```
>>> "Hello!"
'Hello!'

>>> 'how are you?'
'how are you?'

>>> 'short- and long-term'
'short- and long-term'
```

Dictionaries -1

- Lists/Tuples/Strings are **ordered** collections
- Dictionaries are **mapped** (unordered) collections
  - data is accessed using keys
  - data is unordered

- Example:

my_dictionary = {key_1:data_1, key_2:data_2,... , key_n:data_n}

key_1 maps to *data_1*, *key_2* maps to *data_2, etc.*

my_dict = {'Name': 'Jane Smith', 'SN': 990632802, 'Status: 'Registered'}

Dictionaries - Properties

- A dictionary keeps track of associations between keys and values

  - like a regular dictionary where a ***key*** is a word and the

    value is the definition of that word.

    ► This makes look-ups and other operations very easy

- Dictionary values can be any Python object (standard objects or user-defined objects.

- Dictionary keys must be immutable objects: strings, numbers, tuples, etc.:

  - no duplicate key is allowed

Dictionary Functions

- *cmp(dict1, dict2)* Compares elements of both dict
- *len(dict)* Gives the total length of the dictionary, i.e., the number of items in the dictionary
- *str(dict)* Produces a string representation of a dictionary

Dictionary Methods

- **clear()** Removes all elements of the dictionary
- **copy()** Returns a shallow copy of the dictionary
- **fromkeys()** Create a new dictionary with keys from seq and values set to value.
- **get(key, default=None)** Returns the value key maps to or default if key is not in the dictionary
- **has_key(key)** Returns true if key is in the dictionary, false otherwise
- **items()** Returns the data in the dictionary as a list of tuples (key, value)
- **keys()** Returns the keys in the dictionary as a list
- **update(dict2**) Adds dict2's key-values pairs
- **values()** Returns the values in the dictionary as a list

- Form to define a dictionary:

```
my_dict = {key1: value1, key2: value2, ...}
```

- Form to look up a key's value:

```
my_dict[key]
```

- Form to add or update a key-value pair:

```
my_dict[key] = value
```

- Form to delete from a dictionary

```
del dict['Name']    # remove entry with key 'Name'
dict.clear()        # remove all entries in dict
del dict            # delete entire dictionary
```

- **my_dict** is the name of the dictionary object.
- **key** is any immutable object (string, int, tuple).
- **value** is any Python object.

Example -1

```
>>> CO2_by_year = {1799:1, 1800:70, 1801:74,
... 1802:82, 1902:215630, 2002:1733297}

>>> # Look up the emissions for the given year

>>> CO2_by_year[1801]
74

>>> # Add another year to the dictionary

>>> CO2_by_year[1950] = 734914

>>> CO2_by_year{1799: 1, 1800: 70, 1801: 74,
... 1802: 82, 1902: 215630, 2002: 1733297,
... 1950: 734914}
```

Example -2

```
>>> CO2_ by_year[2009] = 1000000
>>> CO2_by_year[2000] = 10
>>> CO2_by_year
{2000: 10, 2002: 1733297, 1799: 1, 1800: 70, 1801: 74, 1802:
82, 2009: 1000000, 1902: 215630}
>>> 1950 in CO2_by_year
False
>>> del CO2_by_year[1950]
>>> len(CO2_by_year) 8
>>> for key in CO2_by_year:
. . .       print(key)
2000
2002
...
```

- How can we iterate through the keys?

```
for k in d.keys():
    print(k)
```

- How can we iterate through the values?

```
for v in d.values():
    print(v)
```

- How can we iterate through the key-value pairs?

```
for key, value in dict.items():
```

Tuples

- Are similar to lists, but cannot be changed

```
my_tuple = (1, 2, 3)
```

- Useful when we want to group data together, in assignment statements

Sets

- Like mathematical sets, they are unordered!

```
>>> my_set = {1, 2, 3}
```

```
>>> x = [1, 2, 3]
>>> my_set = set(x)
```

- An element is either in the set or it is not.

→ efficient membership check

Assignment statements - 1

- General form of an assignment statement: **variable = expression**
- General rule for executing an assignment statement:

1. Evaluate the expression to the right of the = sign.
 -produces a memory address of the value the expression evaluates to

2. Store the memory address in the variable on the left of the = sign.

Assignment statements - 2

```
>>> difference = 20
>>> double = 2 * difference
>>> double
40
>>> difference = 5
>>> double
40
```

- The expression on the right of the = sign is evaluated to 20
- The value 20 will be put at memory address **id1.**
- The variable on the left of the = sign, `difference`, will refer to 20 by storing **id1** in `difference`.

Assignment statements - 3

```
>>> double = 2 * difference
```

- The expression on the right of the = sign: 2 * `difference` `is evaluated`

  => `difference` refers to the value 20

  => `difference` * 20 evaluates to 40.

- The memory address id2 is assigned to the value 40.

- The variable on the left of the = sign, `double`, will refer to 40 by storing id2.

Assignment statements - 4

```
>>> base = 20
>>> height = 12
>>> area = base * height / 2
>>> area
>>> 120.0
>>> celsius = 22
>>> fahrenheit = celsius * 9/5 + 32
>>> fahrenheit
71.6
```

Assignment statements - 5

Strings can also be stored as variables.

```
>>> reminder_text = 'Buy groceries after work'
>>> reminder_text
'Buy groceries after work'
>>> str_var = 'Welcome to MIE1624'
>>> str_var
'Welcome to MIE1624'
>>>str_var = "What is 10 * (2 + 9)?"
>>> str_var
'What is 10 * (2 + 9)?'
```

Assignment statements - 6

Strings can also be stored as variables.

```
>>> reminder_text = 'Please buy groceries after
work'
>>> reminder_text
'Please buy groceries after work'
>>>  str_var  =  'Welcome  to  MIE1624'
>>>  str_var
'Welcome to MIE1624'
>>>str_var = "What is 10 * (2 + 9)?"
>>> str_var
'What is 10 * (2 + 9)?'
```

Augmented Assignment Operators

```
>>> number = 3
>>> Number
3
>>> number  = 2 * number
>>> number
6
>>> number  = number * number
>>> number
36
>>> score  = 50 >>> score 50 >>> score = score +
              20
>>> score 70
```

Augmented Assignment Operators

Operator	Expression	Identical Expression	English description
`+=`	`x = 7` `x += 2`	`x = 7` `x = x + 2`	x refers to 9
`-=`	`x = 7` `x -= 2`	`x = 7` `x = x - 2`	x refers to 5
`*=`	`x = 7` `x *= 2`	`x = 7` `x = x * 2`	x refers to 14
`/=`	`x = 7` `x /= 2`	`x = 7` `x = x / 2`	x refers to 3.5
`//=`	`x = 7` `x //= 2`	`x = 7` `x = x // 2`	x refers to 3
`%=`	`x = 7` `x %= 2`	`x = 7` `x = x % 2`	x refers to 1
`**=`	`x = 7` `x **= 2`	`x = 7` `x = x ** 2`	x refers to 49

What is a function?

- A function is a block of organized (**reusable**) code that is used to perform an activity.

- In Python a function is implemented as a *compound statement*.

- Python has **built-in** functions, but programmers can also create their own **user-defined** functions.

Defining a Python Function -1

- The general form of a function definition:

```
def  function_name (parameters):
     ['''function_docstring''']
     function_body
     [return  [expression]]
```

- A **function block** begins with the keyword **def** followed by the function name and parentheses ( ).

- **Parameters/arguments**:

-0 or more, separated by a comma, are placed within the parentheses.
-variables whose values are supplied when the function is called
-by default, parameters have a positional behaviour (exception: named arguments, which can be given in any order)

Defining a Python Function -2

Function body: consists of one or more statements,

- The code block/body within every function starts with a colon (:) and is *indented*.
- The first statement of a function can be an optional statement - the documentation string of the function, a.k.a. the docstring.
- The statement **`return[expression]`** exits if a function is passing back a value to the caller.

  => A return statement with no arguments is the same as **`return None`**.

Using Functions

- Defining a function only gives the function a name, declares its input parameters and specifies its behaviour, i.e., the instructions to be performed when the function is executed => **but nothing gets executed yet!**

- Once the definition of a function is ready, the function can be executed by **calling** it from another function or directly from the Python prompt.

Calling a Function

- The general form of a function call:

```
function_name(arguments)
```

- Executing a function call:
  - > Evaluate the arguments
  - > Call the function, passing in the argument values
    - >> the instructions in the body of the function are carried out

Function Design Recipe - Six Steps

1. Pick a meaningful name: a short answer to 'What does the function do'?

2. Prepare the Type Contract and write the function header

   > What are the parameter types?
   > What type of value is returned?

   - *Pick meaningful parameter names: it is much easier to understand a function if the variables have names that reflect their meaning.*

3. Prepare a few examples (function calls) - 'What should the function do'?

4. Description: write a docstring describing the function. Mention every parameter in your description. Describe the return value.

5. Body - Write the body of the function.

6. Test - Run the examples designed in Step 3 to make sure they work as expected.

Applying the Design Recipe -1

The United States measures temperature in Fahrenheit and Canada measures it in Celsius. When travelling between the two countries it helps to have a conversion function. Write a function that converts from Fahrenheit to Celsius.

1. **Pick a name**: convert_to_celsius

2. **Type Contract and Header** (what the function will look like)

Type Contract `(number) -> number`

Header `def convert_to_celsius(fahrenheit):`

3. Examples

convert_to_celsius(32) => 0

4.Description `Return the number of Celsius degrees equivalent to Fahrenheit degrees.`

5. **Body**

```
degrees = (fahrenheit - 32) * 5 / 9
return degrees
```

Applying the Design Recipe -2

Complete function definition

```
def convert_to_celsius(fahrenheit):
    ''' (number) -> number
    Return the celsius degrees equivalent to
    fahrenheit degrees.
    '''
    celsius = (fahrenheit - 32) * 5 / 9

    return celsius
```

6. Test - run the examples.

```
>>> convert_to_celsius(32)
0
>>> convert_to_celsius(212)
100
```

Calling functions within other function definitions -1

Let us write a function to convert from hours to seconds.

```
def convert_to_minutes(num_hours):
"""(number) -> number
Return the number of minutes there are in num_hours
hours.
"""
    result = num_hours * 60
 return result

def convert_to_seconds(num_hours):
 """(number) -> number
Return the number of seconds there are in num_hours
hours.
"""
    return convert_to_minutes(num_hours) * 60
```

Testing:

```
>>> convert_to_minutes(2)
120
>>>convert_to_seconds(2)
7200
```

Calling functions within other function definitions -2

```
def convert_to_celsius(fahrenheit):
    '''  (number)   ->  number
    Return  the  number  of celsius   degrees
    equivalent   to  fahrenheit   degrees.  '''

    degrees =   (fahrenheit   - 32)  * 5  /  9
    return  degrees

def convert_to_kelvin(fahrenheit):
    '''  (number)   ->  number
    Return  the  number  of kelvin   degrees equivalent to
    fahrenheit degrees.
    '''
    kelvin  = convert_to_celsius(fahrenheit) + 273.15
    return  kelvin
```

Testing:

```
>>> convert_to_kelvin(32)
273.15
```

Use Function Calls as Arguments to Other Functions

One triangle has a base of length 3.8 and a height of length 7.0 and a second triangle has a base of length 3.5 and a height of length 6.8. Find the area of the larger triangle.

The approach: pass calls to function area as arguments to built-in function max.

```
>>> max(triangle_area(3.8, 7.0), triangle_area(3.5, 6.8))
```

Modules

- A module is a file containing Python definitions and statements.
- The file name is the module name with the suffix .py appended.
- Example: fibo.py module

```
# Fibonacci numbers module
def fib(n):    # write Fibonacci series up to n
    a, b = 0, 1
    while b < n:
        print b,
        a, b = b, a+b
def fib2(n):   # return Fibonacci series up to n
    result = []
    a, b = 0, 1
    while b < n:
        result.append(b)
        a, b = b, a+b
return result
```

- When needed just: import fibo

Opening Files

- **`open(filename, mode)`**

  (str,str) -> io.TextIOWrapper

  opens the f le Filename

  in the same directory as the .py file

  returns a f ile-handle

  mode can take several values:

  r: open the f le for reading

  w: open the f ile for writing (erasing the content!)

  a: open the f le for writing, appending new

  information to the end of the file

Opening Files -2

- To start using a file, given its filename , it has to be open. (The name is a string.)
- To open the file, use the function `open()`

```
myfile = open("story.txt", "r")
```

  - **`open()`** is a Python function
  - **`story.txt`** is the name of the file to be open
  - **`myfile`** is a variable that is assigned the file object returned by open
  - **`r`** is a string indicating what we wish to do with the file. Options for this string are "r", "w", "a", meaning read, write or append. The default is "r "

Note: writing to a file that already exists, erases the existing content. Use append if you want to preserve the content.

Closing Files

```
myfile.close()
```

→ `(NoneType) -> NoneType`

→ myf ile is the f ile object returned by open()

Reading Files

- We call a file object that was opened for reading a **reader**

- Various ways to read from a reader:

1. Read lines one at a time from beginning to end:

   for line in myfile:

   <statements>

2.Read everything in the file at once into a list of strings: read the whole file into list str_ls. Each element of str_ls is a line.

str_ls = myfile.readlines()

print(str_ls)

3. Read everything in the file at once into a string:

```
s = myfile.read() # Read the whole file into string s.
print(s)
```

4. Read a certain number of characters:

```
s = myfile.read(10) # Read 10 characters into s
print(s)
```

5. Read a line at a time:

```
s = myfile.readline() # Read a line into s.
print(s)
s = myfile.readline() # Read the next line into s.
print(s)
```

Reading Files - Recap

- **`myfile.readline()`** – read 1 line from the file
- **`myfile.read()`** - read the whole file into a single string
- **`myfile`.readlines()** - read the whole file into a list, with each element being one line of text
- **`myfile.readlines(n)`** - read the next N bytes of a file, rounded up to the end of a line.

Reading CSV in Python

```
import csv

...

input_file = open(file_name)

reader = csv.reader(input_file)

...

for line in reader:

 # Read as from an ordinary file, but line
is a list

   <process line>
```

Writing to a file

- First we open a f le to write, then we write the contents

  filehandle.write()

Just like printing, except you have to add your own newline characters

- Close your file

Introduction to Pandas

Lecture outline

Introduction to Pandas

- Introduction to pandas data structures – DataFrame, index objects
- Pandas essential functionality
- Summarizing and computing descriptive statistics
- Pivot tables in pandas

Web-scrapping with Python

Introduction to Pandas

- an open source Python library providing high performance data structures and analysis tools.

```
>>>  import pandas as pd

>>>  import numpy as np

>>>  import matplotlib.pyplot as plt
```

Pandas Data Structures -Series

- One-dimensional labeled array
- Holds any data type (integers, strings, floating point numbers, Python objects, etc.)
- The axis labels are collectively referred to as the index.

```
>>> s = pd.Series(data, index=index)
```

- **data**: a dictionary, an ndarray, a scalar value (e.g., 11)
- **index:** is a list of axis labels.

Series from from an ndarray

```
>>> s= pd.Series(np.random.randn(5), index=['a',
'b','c',  'd', 'e'])
>>> s
a    0.2735
b    0.6052
c   -0.1692
d    1.8298
e    0.5432
dtype: float64

>>> s.index
Index(['a', 'b', 'c', 'd', 'e'], dtype='object')

>>> pd.Series(np.random.randn(5))
0    0.3674
1   -0.8230
2   -1.0295
3   -1.0523
4   -0.8502
dtype: float64
```

Series from from a dictionary

- If an **index** is passed, the values in data corresponding to the labels in the index will be pulled out.
- If no **index** is passed, an index will be constructed from the sorted keys of the dict, if possible.

```
>>> d = {'a' : 0., 'b' : 1., 'c' : 2.}
>>> pd.Series(d)
a    0.0
b    1.0
c    2.0
dtype: float64
>>> pd.Series(d, index=['b', 'c', 'd', 'a']) b 1.0
c    2.0
d    NaN
a    0.0
dtype: float64
```

- **NOTE**: NaN is the standard missing data marker used in pandas

Series from from a scalar value

- If **data** is a scalar value, an **index** must be provided. The value will be repeated to match the length of index

```
>>> pd.Series(5., index=['a', 'b', 'c', 'd', 'e'])
a    5.0
b    5.0
c    5.0
d    5.0
e    5.0
dtype: float64
```

Series Behaviour

- Series acts very similarly to a ndarray, and is a valid argument to most NumPy functions.

```
>>> s[0]
>>> 0.27348116325673794
>>> s[:3]
>>> a    0.2735
    b    0.6052
    c   -0.1692
    dtype: float64
```

- A Series is like a fixed-size dict in that you can get and set values by index label:

```
>>> s['a']
>>> 0.27348116325673794
>>> s['e'] = 12.

>>> s.get('a')
>>> 0.27348116325673794
```

DataFrame Objects

- class pandas.DataFrame(data=None, index=None, columns=None, dtype=None, copy=False)[source]

- Two-dimensional, size-mutable, potentially heterogeneous tabular data structure with labeled rows and columns.

- Dictionar-like container for Series objects.

- Arithmetic operations align on both row and column labels.

DataFrame - Parameters

- **data** : numpy ndarray, dictionary (Series, arrays, constants, or list-like objects), or DataFrame
- **index** : index or array-like to use for resulting frame.
- **columns** : Index or array-like, labels to use for resulting frame.
- **dtype** : data_type (default None), to force, otherwise infer
- **copy** : boolean (default False), to copy data from inputs.

```
>>> d = {'col1': ts1, 'col2': ts2}
>>> df1 = DataFrame(data = d, index = index)
>>> df2 = DataFrame(numpy.random.randn(10, 5))
>>> df3 = DataFrame(numpy.random.randn(10, 5),
columns=['a', 'b', 'c', 'd', 'e'])
```

- numpy.random.randn returns a sample(s) from the "standard normal" distribution.

DataFrames from Series or dictionaries

- The result index will be the union of the indexes of the various Series.

```
d = {'one' : pd.Series([1., 2., 3.], index=['a', 'b', 'c']),
     'two' : pd.Series([1., 2., 3., 4.], index=['a', 'b', 'c', 'd'])}
>>> df = pd.DataFrame(d)
>>> df
one  two
a  1.0  1.0
b  2.0  2.0
c  3.0  3.0
d  NaN  4.0

>>> pd.DataFrame(d, index=['d', 'b', 'a'])
one  two
d  NaN  4.0
b  2.0  2.0
a  1.0  1.0
```

Accessing Rows and Columns

- The row and column labels can be accessed, respectively, by accessing the index and columns attributes:
- **Note**: when a particular set of columns is passed along with a dict of data, the passed columns override the keys in the dict.

```
>>> df.index
Index([u'a', u'b', u'c', u'd'], dtype='object')
>>> df.columns
Index([u'one', u'two'], dtype='object')
```

Index

- Immutable ndarray implementing an ordered, sliceable set. The basic object storing axis labels for all pandas objects
- Parameters:
- data : array-like (1-dimensional)
- dtype : NumPy dtype (default: object)
- copy : bool Make a copy of input ndarray
- name : objectName to be stored in the index
- tupleize_cols : bool (default: True)

  When True, attempt to create a MultiIndex if possible

Index Attributes

Attributes

`T`	return the transpose, which is by definition self
`asi8`	
`base`	return the base object if the memory of the underlying data is
`data`	return the data pointer of the underlying data
`dtype`	
`dtype_str`	
`flags`	
`has_duplicates`	
`hasnans`	
`inferred_type`	
`is_all_dates`	
`is_monotonic`	alias for is_monotonic_increasing (deprecated)
`is_monotonic_decreasing`	return if the index is monotonic decreasing (only equal or
`is_monotonic_increasing`	return if the index is monotonic increasing (only equal or
`is_unique`	
`itemsize`	return the size of the dtype of the item of the underlying data

Index Methods

Methods

all(*args, **kwargs)	Return whether all elements are True
any(*args, **kwargs)	Return whether any element is True
append(other)	Append a collection of Index options together
argmax([axis])	return a ndarray of the maximum argument indexer
argmin([axis])	return a ndarray of the minimum argument indexer
argsort(*args, **kwargs)	Returns the indices that would sort the index and its underlying data.
asof(label)	For a sorted index, return the most recent label up to and including the passed label.
asof_locs(where, mask)	where : array of timestamps
astype(dtype[, copy])	Create an Index with values cast to dtypes.
copy([name, deep, dtype])	Make a copy of this object.
delete(loc)	Make new Index with passed location(-s) deleted
difference(other)	Return a new Index with elements from the index that are not in *other*.
drop(labels[, errors])	Make new Index with passed list of labels deleted
drop_duplicates(*args, **kwargs)	Return Index with duplicate values removed

Reshaping by pivoting DataFrame objects -1

- Reshaping by pivoting DataFrame objects
- Data is often stored in CSV files or databases in so-called "stacked" or "record" format:

```
>>> df
```

	date	variable	value
0	2000-01-03	A	0.469112
1	2000-01-04	A	-0.282863
2	2000-01-05	A	-1.509059
3	2000-01-03	B	-1.135632
4	2000-01-04	B	1.212112
5	2000-01-05	B	-0.173215
6	2000-01-03	C	0.119209
7	2000-01-04	C	-1.044236
8	2000-01-05	C	-0.861849
9	2000-01-03	D	-2.104569
10	2000-01-04	D	-0.494929
11	2000-01-05	D	1.071804

Reshaping by pivoting DataFrame objects -2

- To select out everything for variable A we could do:

```
>>> df[df['variable'] == 'A']
>>> date           variable    value
0 2000-01-03       A           0.469112
1 2000-01-04       A           -0.282863
2 2000-01-05       A           -1.509059
```

- For time series operations a better representation would have the columns as unique variables and an index of dates identifying individual observations.
- To reshape the data use the **pivot** function:

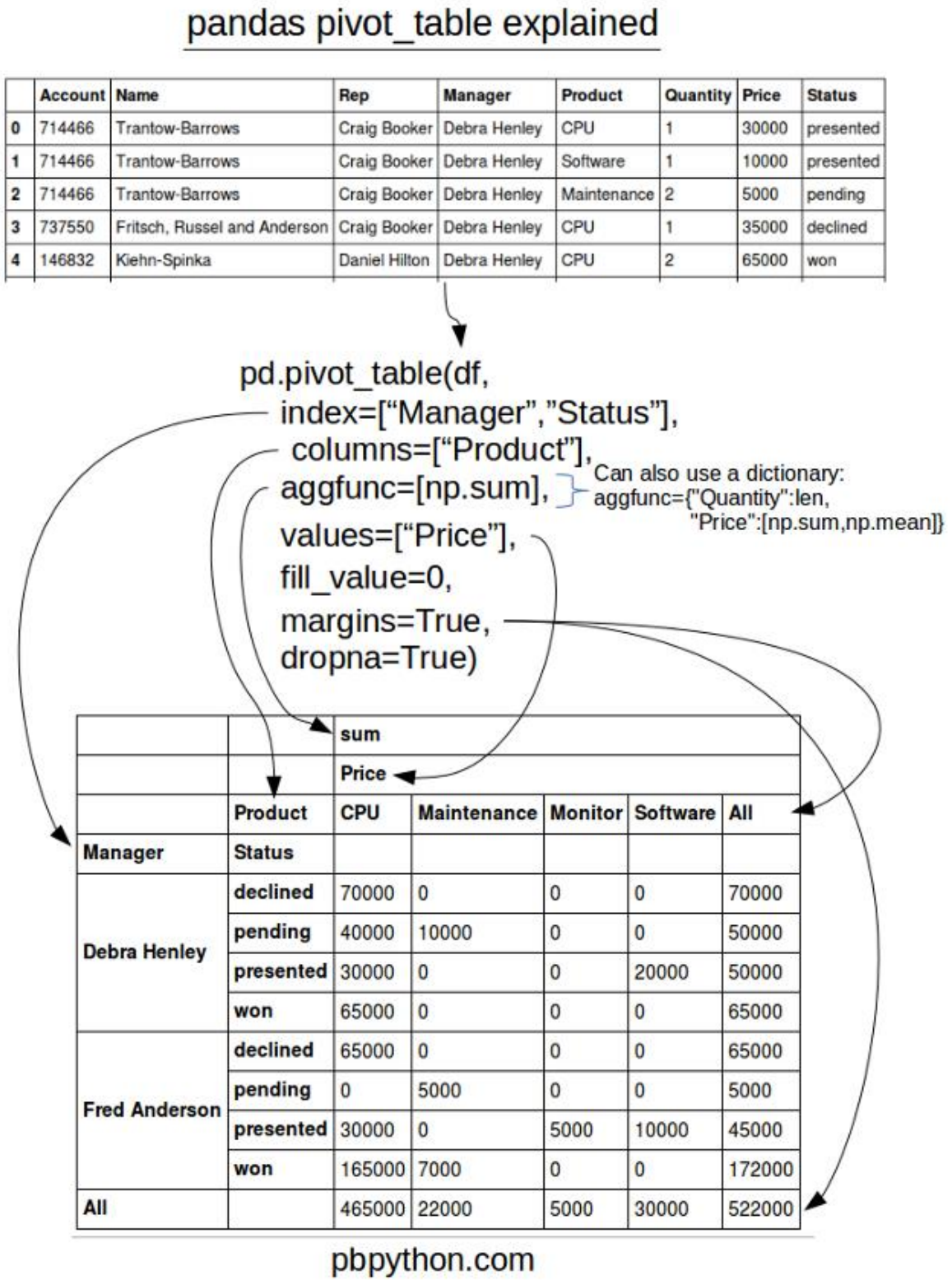
```
>>> df.pivot(index='date', columns='variable', values='value')
>>> variable              A          B          C          D
    date
    2000-01-03    0.469112  -1.135632   0.119209  -2.104569
    2000-01-04   -0.282863   1.212112  -1.044236  -0.494929
    2000-01-05   -1.509059  -0.173215  -0.861849   1.071804
```

Computing Descriptive Statistics

- DataFrame.describe(percentiles=None, include=None, exclude=None)[source]
- Generate various summary statistics, excluding NaN values.
- **Parameters**:
- **percentiles** : array-like, optional. The percentiles to include in the output. Should all be in the interval [0, 1].
- **include**, **exclude** : list-like, 'all', or None (default) Specify the form of the returned result. Either:
  - None to both (default). The result will include only numeric-typed columns or, if none are, only categorical columns.
  - A list of dtypes or strings to be included/excluded.
  - If **include**= 'all', the output column-set will match the input one.

Returns: summary statistics

Pandas pivot_table cheat sheet



	Account	Name	Rep	Manager	Product	Quantity	Price	Status
0	714466	Trantow-Barrows	Craig Booker	Debra Henley	CPU	1	30000	presented
1	714466	Trantow-Barrows	Craig Booker	Debra Henley	Software	1	10000	presented
2	714466	Trantow-Barrows	Craig Booker	Debra Henley	Maintenance	2	5000	pending
3	737550	Fritsch, Russel and Anderson	Craig Booker	Debra Henley	CPU	1	35000	declined
4	146832	Kiehn-Spinka	Daniel Hilton	Debra Henley	CPU	2	65000	won

		sum				
		Price				
	Product	CPU	Maintenance	Monitor	Software	All
Manager	Status					
Debra Henley	declined	70000	0	0	0	70000
	pending	40000	10000	0	0	50000
	presented	30000	0	0	20000	50000
	won	65000	0	0	0	65000
Fred Anderson	declined	65000	0	0	0	65000
	pending	0	5000	0	0	5000
	presented	30000	0	5000	10000	45000
	won	165000	7000	0	0	172000
All		465000	22000	5000	30000	522000

Jupyter Notebooks

Jupyter Notebook

- An *interactive computational environment*, in which you can combine code execution, rich text, mathematics, plots and multi media.

- *Notebook documents* (“notebooks”) are documents produced by the Jupyter Notebook App

  contain: computer code (e.g. python) and rich text elements (paragraph, equations, figures, links, etc.).

- Notebook documents are both human-readable documents as well as executable documents which can be run to perform data analysis.

Jupyter Notebook App

- A server-client application that allows editing and running notebook documents via a web browser.

- The Jupyter Notebook App can be executed on a local desktop requiring no internet access or can be installed on a remote server and accessed through the internet.

- In addition to displaying/editing/running notebook documents, the App has a “Dashboard” showing local files and allowing to open notebook documents.

Notebook Kernel

A *computational engine* that executes the code contained in a Notebook document.

- The kernel associated with a notebook is automatically launched when the notebook is opened.

- When the notebook is executed, the kernel performs the computation and produces the results.

Notebook Dashboard

- The Dashboard is the component which is shown first in the Jupyter App.
- Main functionality: open notebook documents, and to manage the running kernels.
- Other features (similar to a file manager): navigating folders and renaming/deleting files.

Running the Jupyter Notebook

- The App can be launched by clicking on the Jupyter Notebook icon installed by Anaconda in the start menu (Windows) or by typing in a terminal (cmd on Windows, Terminal on OSX):

  > jupyter notebook

- This will launch a new browser window/tab showing the Notebook Dashboard
- When started, the App can access only files within its start-up folder (including any sub-folder).
- If the notebook documents are not in a subfolder of your user folder further configuration steps are necessary.

Shutting Down/Restarting a Kernel

- When a notebook is opened, its kernel is also started.
- Closing the notebook browser tab, will not shut down the kernel => the kernel needs to be explicitly shut down.
- To **shut down** a kernel:

  (1) go to the associated notebook and click on menu **File** ->**Close and Halt.**

  or (2) in the **Running** tab of the **Dashboard** (which shows all the running notebooks/kernels) click on a the **Shutdown** button of the kernel you want to stop.

- To **restart** a kernel click on the menu **Kernel -> Restart.** This can be useful to start over a computation from scratch

Executing a Notebook

- Launch the App
- In the Dashboard, navigate to find the notebook.
- Click on its name (will open it in a new browser tab).
- Run the notebook step-by-step (one cell a time) by pressing **shift + enter**.
- Run a notebook in a single step by clicking on the menu **Cell -> Run All**.

Shut down the Jupyter Notebook App

- Closing the browser window/tab will not close the Jupyter Notebook App.

  => to completely shut it down the associated terminal needs to be closed.

- Many copies of the Jupyter Notebook App can be run in parallel, but it is not a recommended usage mode.

📖 To Do before Lecture 3

Run IPython examples provided in class

- **Use Python on cloud via Google Colab**
  - You can use Python on Google cloud via https://colab.research.google.com
- **Install Python on your laptop**
  - Recommended to use Python version 3.9, 3.10 or 3.11
  - You may use your own Python distribution, Anaconda distribution is recommended to install https://www.anaconda.com/download
- **Form groups of seven students for in-class presentations and course project**
  - Add all your group members to Group X on Quercus
  - All groups should have exactly seven members
  - In-class presentations will be done in the order of group numbers
  - Course Project will be the same for all groups
  - Every group member get the same mark, independently on how you split responsibilities inside each group
- **Check class web-page on Quercus regularly**