



# Introduction to Python

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# Plan for this lecture

- Introduction to Python
- Main goal is to present you a subset of the language and libraries to make you able to tackle Machine Learning challenges with Python.

# Overview

## ■ What is Python?

- ◆ Python is an open general purpose language that is widely used in scientific computing and machine learning.
- ◆ Rich ecosystem of libraries for scientific computation. NumPy for linear algebra, scikit-learn for general Machine Learning, Apache Spark for distributed ML and so on.

# Python

- Python is dynamically typed, that means that the type of an expression is unknown before evaluation time. (but there are types!).
- Supports object-orientation and functional programming.
- We use Python 3 in the lecture.

# Python Basics I

## ■ Hello World:

```
print("Hello, World!")
```

## ■ Variables:

```
x = 5
print(x)
print(type(x))
print("x = " + x)          # does not work
print("x = " + str(x))
```

## ■ Arithmetic:

```
x = 0.5
y = 3
print(y**2 + x)           #  $y^2 + x$ 
print(y / x)
z = 5
print(y / z)
print(float(y) / z)
print(int(x))
```

# Python Basics II

## ■ Boolean algebra:

```
winter = True
rain = False
snow = winter and rain
print(snow)
summer = not winter
print(summer)
bad_weather = winter or rain
print(bad_weather)
```

## ■ Comparison operators:

```
print(5 == 3)           # note that = is an assignment
print(3 < 4)
print(2+2 == 5 and True)
print(4 % 3 == 0)       # a % b is remainder of integer
                        # division of a by b
```

# Python Basics III

- Functions (notice the indentation!):

```
def square_plus(x, y):  
    print("square with x= " + str(x) + " evaluated")  
    return x*x + y  
print(square_plus(3, 1))
```

- Call-by-???:

```
def set_to(x, value):  
    print("x set to " + str(value))  
    x = value
```

```
y = 5  
set_to(y, square_plus(2, 0)) # what happens in  
                             # which order here  
                             # did the original y change?  
  
print y
```

# Python Basics IV

- If (run specific code only if a condition is met):

```
def abs(x):  
    if x < 0:  
        return -x  
    return x  
print(abs(3), abs(-5))
```

- While (run code while some condition is met prior to each run)

```
i = 4  
while i >= 0:  
    print(i)  
    i -= 1                # equal to i = i - 1
```

- For (run code for each object in an ordered sequence as a parameter)

```
for i in range(5):  
    print(i)  
for c in "WOW!":  
    print(c)
```



# Python Basics V

- Recursion (calling itself-itself-itself-...):

```
def factorial(n):  
    if n == 0:  
        return 1  
    else:  
        return n * factorial(n-1)  
  
print(factorial(42))
```

# Python Basics VI

- A test: find a good name for the following function

```
def what_am_i(n):  
    i = 0  
  
    while i < n:  
        s = ""  
  
        for j in range(n):  
            if j == i or j == n-i-1:  
                s += "*"   
            else:  
                s += " "  
        print(s)  
        i += 1
```

# Python Basics VII

- **Modules: Every file (or directory with `_init_.py`)**

```
import math
```

```
print(math.sin(3))  
print(math.factorial(10))
```

```
from math import sin, cos, exp  
print(sin(3)**2 + cos(3)**2)  
print(exp(1))
```

- **Other module: random numbers**

```
from random import *
```

```
print(random())           # 0 to 1 uniform  
print(randrange(10))      # integer 0, 1, ..., 9  
print(uniform(-0.5, 0.5))  
print(gauss(0, 1.0))      # normal distribution
```

- **Standard modules:**

```
collections, string, itertools, os, sys
```

# Python Basics VIII (Data Structures I)

## ■ Lists:

```
xs = [1, 2, 3, 4]
print(xs[0])
print(xs)
print(len(xs), sum(xs))
print([1, 2] + [3]*2 + [])
print("ab" * 3)
```

## ■ Mutability of lists:

```
some_objects = []
some_objects.append("a")
some_objects.append(True)
some_objects.append(3)
```

```
print(some_objects[0])
print(some_objects)
```

```
del some_objects[0]
some_objects.remove(3)
```

# Python Basics IX (Data Structures II)

## ■ Slicing

```
nums = range(20)
print(nums[1:10])
print(nums[:10])
print(nums[5:])
print(nums[:])
print(nums[10::-1])
print(nums[:10:-1])
print(nums[3:15:4])
```

## ■ List comprehensions

```
squares = [x**2 for x in range(10)]
print(squares)

pythagorean_triples = [(x, y, z) for x in range(1, 10)
                        for y in range(1, 10)
                        for z in range(10) if x**2 + y**2 = z**2 and x > y]
print(pythagorean_triples)
```

# Python Basics X (Data Structures III)

- Dictionaries (hash maps):

```
dictionary = {"Eins": 1, "Zwei": 2, "Drei": 3}
print("Eins" in dictionary)
print 1 in dictionary
del dictionary["Eins"]
```

```
for key, value in dictionary.items():
    print("{}: {}".format(key, value))
```

- Lambda expressions

```
squaring = lambda x: x**2
print(squaring(3))
```

# When you are stuck

- `help` opens documentation.
- `doc(obj)` or `obj?` for any object `obj` (commands, classes, modules)
- `who`, `whos`: lists all currently available identifiers, latter with more detail.
- `del x`: deletes `x` from memory.
- `clear`: clears output if you run Python in a terminal.

# NumPy I

Start with `import numpy as np`

- Input of numbers:

```
>> a=2
2
>> a = np.sqrt(-16 + 0j)
4j
```

- With print explicit display of value:

```
>> print(a)
4j
```

- Or simply writing the name as last expression:

```
>> a
4j
```



# NumPy II

- Defining a vector:

```
>> b = np.ndarray([2, 4, 6, 8])  
[2 4 6 8]
```

This is a vector of length 4 (implicitly row vector)

```
>> b2 = b.reshape(4, 1)  
>> print b.dot(b2)  
array([120])
```

The data lies flat (i.e. sequentially) in memory,  
shape returns logical structure

```
>> print(b.shape, b2.shape)  
(4,) (4, 1)
```

Shapes can be in any higher dimensions, ndarrays  
are in fact tensors.

# NumPy III

- Generate c equidistant points from interval [a, b]:

```
>> b2 = np.linspace(1, 3, 5)
array([ 1., 1.5, 2., 2.5, 3. ])
```

- Generate range as a vector:

```
>> b3 = np.arange(0, 10, 2)
array([0, 2, 4, 6, 8])
```

# NumPy IV

- Input of Matrices:

```
>> A = np.ndarray(np.mat('[1 2 3; 4 5 6; 7 8 0]'))  
array([[1, 2, 3],  
       [4, 5, 6],  
       [7, 8, 0]])
```

This results in a 3x3 matrix.

- Transpose:

```
>> A2 = A.T  
array([[1, 4, 7],  
       [2, 5, 8],  
       [3, 6, 0]])
```

# NumPy V

- Linear Indexing:

```
>> A[0]  
array([1, 2, 3])
```

- Indexing over row and column:

```
>> A[1, 2]      returns 6, zero-based (row, column)
```

- Indexing via lists and slicing:

```
>> A([0, 2], 1)  returns [2, 8]  
>> A[2, :]      returns 3rd row as slice  
>> A[:, 2]      returns 3rd column as slice
```

# NumPy VI

- Change values via assignment:

```
>> A[2,2] = 9
```

```
A =
```

1	2	3
4	5	6
7	8	9

- Matrix shape can be adjusted by reshape, but should not. Create new matrices by operators and creators.
- Information about matrices

`A.shape`    Dimensions, returns (3, 3) here

`A.dtype`    Kind of scalars the matrix contains, i.e. int64, float64

# NumPy VII

- Commands to create matrices:

`np.zeros((n,m))`

nxm matrix with only zeros

`np.ones((n,m))`

nxm matrix with only ones

`np.full((n,m),c)`

nxm matrix with only c

`np.eye(n)`

nxn identity matrix

- Random sampling (more at [SciPy docs](#))

`from np.random import rand, randn`

`rand(n,m)`

nxm matrix with uniform picked entries in the half-open unit interval [0, 1)

`randn(n,m)`

nxm matrix with normally distributed entries (zero mean, unit std)

# NumPy VIII

- Some constants

<code>np.pi</code>	3.14159...
<code>0.+1j</code>	imaginary unit
<code>np.inf</code>	infinity
<code>np.nan</code>	“not a number”

# NumPy IX

## ■ Matrix operators:

<code>+</code>	addition
<code>-</code>	subtraction
<code>np.dot</code>	matrix multiplication
<code>^</code>	matrix exponentiation
<code>np.linalg.solve</code>	left division
<code>.T</code>	transpose
<code>.H</code>	complex-conjugated transpose

## ■ Element-wise operators:

<code>*</code>	element-wise multiplication
<code>**</code>	element-wise exponentiation
<code>/</code>	element-wise division



# NumPy X (Examples)

```
>> x = np.ndarray([-1, 0, 2])  
      array([-1, 0, 2])  
  
>> y = x - 1  
      array([-2, -1, 1])  
  
>> x.T.dot(y)  
      4  
  
>> x.dot(y.T)  
      [adjusted output]  
           2      1      -1  
           0      0      0  
          -4     -2      2  
  
>> y.dot(x.T)  
      [adjusted output]  
           2      0     -4  
           1      0     -2  
          -1      0      2  
  
>> np.pi * x  
      array([-3.1416, 0., 6.2832])
```

# Sources

- <https://continuum.io> – Anaconda distribution, easy to use installation of Python. Works well under Windows. This is also installed for you on the computer lab's Linux systems.
- <http://learnpythonthehardway.org> – A gentle introduction to Python as a general-purpose language.
- <https://www.edx.org> – Decent (and free) online classes for Python.
  - ◆ 6.00.2x: Python intro with scientific/statistical approach. If you lack CS fundamentals start with 6.00.1x.
  - ◆ CS190-1x: Large scale ML with Python and Spark. Labs very cool (e.g. visualization of neuroimage data of Jellyfishes).

# More Sources

- <https://github.com/amueller> – Wonderful collection of tutorials for ML with Python with notebooks, you can find accompanying videos often.
- [https://github.com/parallel\\_ml\\_tutorial](https://github.com/parallel_ml_tutorial) -- Parallel ML with Python. Useful for quicker prototyping.