

# Python102

Python for Data Science Bootcamp

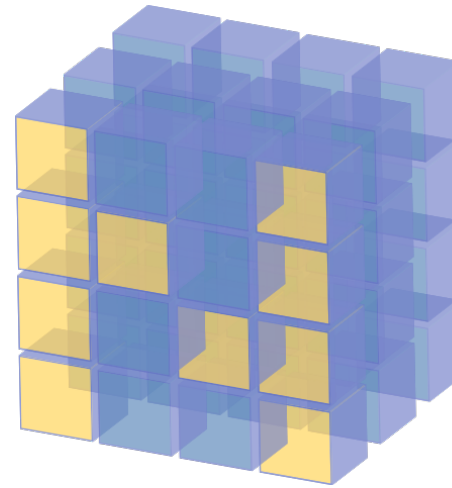
## **(4.1) Python for Data Analysis**

*NumPy*

*AIAT Academy*

# Python for Data Analysis Outline

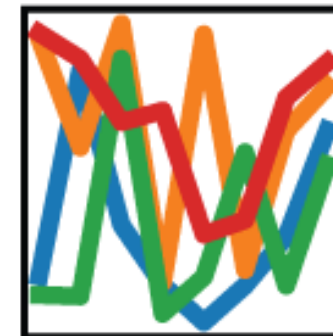
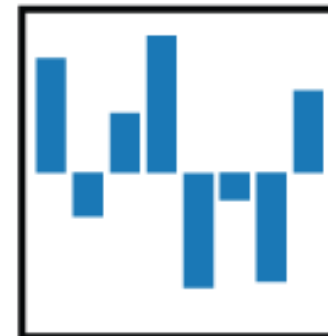
- NumPy
- Pandas



NumPy

pandas

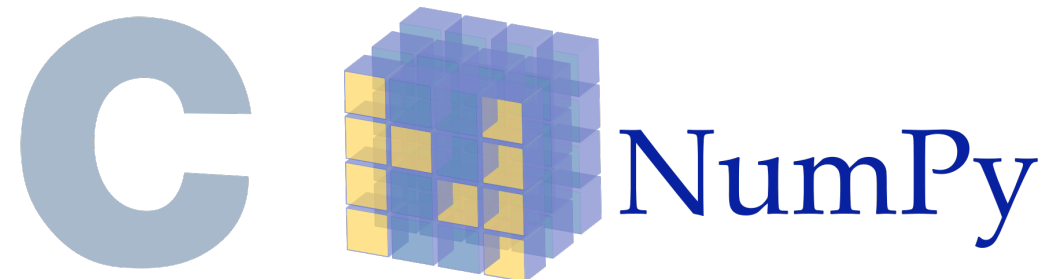
$$y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$$



# NumPy

# Introduction to NumPy

- Fundamental package for scientific computing with Python.
  - Linear Algebra, Fourier transform, etc.
- NumPy is incredibly fast since it has bindings to C libraries.
- Optimized library for matrix and vector computation.
- N-dimensional array object
- Open source



# NumPy Installation

- It is highly recommend you install Python using the Anaconda distribution to make sure all dependencies all sync up with the use of a conda install.
- To install NumPy, just going to your terminal or command prompt and typing

`conda install numpy`

or

`pip install numpy`

# NumPy in Python102

- NumPy arrays are the main way we will use NumPy in this course.
- NumPy arrays come in 2 flavors: Vectors and Matrices.
- Vectors are 1-d arrays and 2-d arrays

Scalar    Vector    Matrix    Tensor

1

$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$

$\begin{bmatrix} \begin{bmatrix} 1 & 2 \end{bmatrix} & \begin{bmatrix} 3 & 2 \end{bmatrix} \\ \begin{bmatrix} 1 & 7 \end{bmatrix} & \begin{bmatrix} 5 & 4 \end{bmatrix} \end{bmatrix}$

# NumPy Arrays

Main data type: **ndarray**  
(**np.ndarray**)

This is the data type that you will use to represent  
vector/matrix computation

Note: Constructor function is **np.array()**



# NumPy Arrays (np.array)

```
import numpy as np
```

```
A = np.array([[1, 2, 3], [4, 5, 6]])
```

```
# >> array([ [1, 2, 3],  
#          [4, 5, 6] ] )
```

```
np.array([Numerical list(s)])
```

# NumPy Arrays (np.array)

```
import numpy as np
```

```
A_float = np.array([1, 2, 3], float)
```

```
# >> array([ [1., 2., 3.],  
#           [4., 5., 6.] ] )
```

# NumPy Arrays (np.arange)

```
import numpy as np
```

```
np.arange(3)           # >> array([0, 1, 2])
```

```
np.arange(3.0)         # >> array([0., 1., 2.])
```

```
np.arange(3, 7)        # >> array([3, 4, 5, 6])
```

```
np.arange(3, 7, 2)     # >> array([3, 5])
```

```
np.arange(start, stop, step)
```

# NumPy Arrays (np.zeros)

```
import numpy as np
```

```
A = np.zeros((2, 3))
```

```
# >> array( [ [0., 0., 0.],
               [0., 0., 0.] ] )
```

```
A.shape
```

```
# >> (2, 3)
```

```
np.zeros(shape)
```

# NumPy Arrays (np.ones)

```
import numpy as np
```

```
A = np.ones((2, 3))
```

```
# >> array( [ [1., 1., 1.],
               [1., 1., 1.] ] )
```

```
A.shape
```

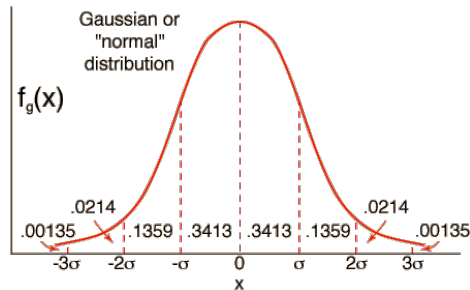
```
# >> (2, 3)
```

```
np.ones(shape)
```

# NumPy Arrays (np.random)

```
np.random.random((2, 3))
```

```
# >> array([[ 0.78084261,  0.64328818,  0.55380341],
             [ 0.24611092,  0.37011213,  0.83313416]])
```



```
np.random.normal(loc=1.0, scale=2.0, size=(2, 2))
```

```
# >> array([[ 2.87799514,  0.6284259 ],
             [ 3.10683164,  2.05324587]])
```

# NumPy Arrays (Array Attributions)

```
a = np.arange(10).reshape((2, 5))
```

<code>a.ndim</code>	# 2 (dimensions)
<code>a.shape</code>	# (2, 5) (shape of array)
<code>a.size</code>	# 10 (number of elements)
<code>a.T</code>	# Transpose
<code>a.dtype</code>	# Data type

# NumPy Indexing and Selection



# NumPy Indexing and Selection

```
a = np.arange(4)      # >> array( [0, 1, 2, 3] )
```

```
a[1]      # >> 1
```

```
a[1:3]    # >> array( [1, 2] )
```

```
a[:3]     # >> array( [0, 1, 2] )
```

```
a[1:]     # >> array( [1, 2, 3] )
```

```
a[1:3] = 10 # >> a = array( [0, 10, 10, 3] )
```

# NumPy Indexing and Selection

## (Nested Arrays | 2D+ Arrays)

```
a_2d = np.array([ [1, 2, 3], [4, 5, 6] ])
```

```
# >> array( [ [1, 2, 3],
               [4, 5, 6] ] )
```

```
a_2d[1]      # >> array([4, 5, 6])
```

```
a_2d[1][1]   # >> 5
```

```
a_2d[1,1]    # >> 5 (same as a_2d[1][1])
```

# NumPy Indexing and Selection

## (Nested Arrays | 2D+ Arrays)

```
a_2d = np.array([ [1,2,3], [4,5,6], [7,8,9] ])
```

```
# >> array( [ [1, 2, 3],
               [4, 5, 6],
               [7, 8, 9] ] )
```

```
a_2d[:2]
```

```
# >> array( [ [1, 2, 3],
               [4, 5, 6] ] )
```

# NumPy Indexing and Selection

## (Nested Arrays | 2D+ Arrays)

```
a_2d = np.array([ [1,2,3], [4,5,6], [7,8,9] ])
```

```
# >> array( [ [1, 2, 3],
               [4, 5, 6],
               [7, 8, 9] ] )
```

```
a_2d[:2, 1:]
```

```
# >> array( [ [2, 3],
               [5, 6] ] )
```

# NumPy Indexing and Selection (Statement)

```
a = np.arange(1,5)
```

```
# >> array([1, 2, 3, 4])
```

```
a > 2
```

```
# >> array([False, False, True, True ], dtype=bool)
```

```
a[a > 2]      # >> array([3, 4])
```

```
a[a < 2]      # >> array([1])
```

# NumPy Operation

# NumPy Arrays (Operations)

```
a = np.arange(4)
```

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

```
b = np.array([2, 3, 2, 4])
```

```
a * b          # array( [ 0, 3, 4, 12] )
```

```
b - a          # array( [ 2, 2, 0, 1] )
```

```
c = [2, 3, 4, 5]
```

```
a * c          # array( [ 0, 3, 8, 15] )
```

# NumPy Arrays (Operations)

```
a = np.arange(4)
```

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

```
b = np.array([2, 3, 2, 4])
```

```
a - 10      # array( [ -10, -9, -8, -7] )
```

```
np.exp(a)   # array([ 1., 2.718, 7.389, 20.085])
```

For more operations:

<https://docs.scipy.org/doc/numpy/reference/ufuncs.html>



# NumPy Arrays (Vector Operations)

# NumPy automatically converts lists

```
u = [1, 2, 3]          v = [1, 1, 1]
```

```
np.inner(u, v)         # 6
```

```
np.outer(u, v)
```

```
# array([[1, 1, 1],
```

```
#      [2, 2, 2],
```

```
#      [3, 3, 3]])
```

```
np.dot(u, v)          # 6
```