

## Lesson2 Using Numpy

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**Numpy** lets you do additional operations on arrays. **Numpy arrays** are similar to Python **lists** but are not the same and have more features. Numpy arrays are made from python **lists**. To use numpy arrays you must import the numpy package at the top of your python file. Numpy stands for numeric python.

```
import numpy as np
```

**as np** means instead of saying **numpy** you can say **np** instead for convenience

recap: numpy arrays are array objects that you create from python lists.

### Create numpy arrays from python lists

#### Create a 1 d numpy array

To create numpy array of 5 elements you would use this statement:

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

Print out numpy array **a**:

```
[ 1 2 3 4 5]
```

Numpy arrays have a **rank**, **shape** and **type**.

<b>rank</b>	number of dimensions	1
<b>shape</b>	size of the array for each dimension return as a tuple (size1, size2, size3)	(5, ) (5 rows no columns)
<b>type</b>	data type like int or float	int, float

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

The above array would have rank of 1 dimensions of shape size (5, ) and data type int. (5, ) means 1 dimension of 5 elements.

```
print(np.rank(a)) # 1  
print(a.shape) # (5,)  
print(a.dtype) # int32
```

### **Create a 2 d numpy array**

We can make a 2 dimensional numpy array from a 2 dimensional python array

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

Print out numpy array **b**:

```
[[1 2 3]  
 [4 5 6]]
```

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

The above array would have rank of 2 dimensions of shape size (2,3) and data type int and 2 rows and 3 columns

```
print(np.rank(b)) # 2  
print(b.shape) # (2,3)  
print(b.dtype) # int32
```

You can use numpy arrays just like python lists

```
a = np.array([1,2,3,4,5]) # [1,2,3,4,5]  
a[0] = 5; # assign 5 to index 0 of numpy array a  
x = a[0]; #retrieve value from number array a at index 0  
print(x) # print out x
```

1

**`x = a[-1]; #retrieve value from array a at last index`**

5

read rows and columns from a 2 dimensional array

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

**`x = b[1][0]; #retrieve value from array row index 1 column index 0`**

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### creating Numpy arrays

Numpy has many function's to create arrays as follows:

**# Create a 2x2 array of all zeros**

**`a = np.zeros((2,2))`**

[[0,0][0,0]]

```
# Create a 2x2 array of all ones  
a = np.ones((2,2))
```

```
[[1,1][1,1]]
```

```
# Create a 2x2 array of a specified value  
a = np.full((2,2), 5)
```

```
[[5,5][5,5]]
```

```
# Create a 2x2 identity matrix  
a = np.eye(2)
```

```
[[1,0][0,1]]
```

```
# Create an array with random values 0 to 1  
a = np.random.random((2,2))
```

```
[[0.5484408 0.57400321],[0.19896352 0.65698084]]
```

```
#create an array with a sequence of 0 to 18 by step of 2  
a = np.arange(0, 20, 2)
```

```
[0, 2, 4, 6, 8,10,12,14,16,18]
```

```
#create an array of even space between a given range of values  
# (divides 0 to 1 evenly by 5)  
a = np.linspace(0, 1, 5)
```

```
([ 0., 0.25, 0.5 , 0.75, 1.]
```

```
#create a 3x3 array with mean 0 and standard deviation 1  
# for a specified array dimension  
a = np.random.normal(0, 1, (3,3))
```

```
[[ 0.72432142, -0.90024075, 0.27363808],  
 [ 0.88426129, 1.45096856, -1.03547109],  
 [-0.42930994, -1.02284441, -1.59753603]]
```

## Concatenating Numpy Arrays together

```
# Concatenate 1 Dimensional arrays together
```

```
x = np.array([1, 2, 3])  
y = np.array([4, 5, 6])  
a = np.concatenate([x, y])
```

```
[1 2 3 4 5 6]
```

**# Concatenate 2 Dimensional arrays together (default: row-wise axis=0)**

```
x = np.array([[1, 2, 3],[4,5,6]])  
y = np.array([[1,2,3],[4,5,6]])  
b = np.concatenate([x, y]) # default axis = 0 (row axis)
```

```
[[1 2 3]  
 [4 5 6]  
 [1 2 3]  
 [4 5 6]]
```

**# Concatenate 2 Dimensional arrays together (column-wise)**

```
x = np.array([[1, 2, 3],[4,5,6]])  
y = np.array([[1,2,3],[4,5,6]])  
b = np.concatenate([x, y],axis=1) # 1 = column axis
```

```
[[1 2 3 1 2 3]  
 [4 5 6 4 5 6]]
```

**vstack (vertical stack)**

**vstack** stacks a 1 d array on the top of a 2 d array vertically.

before:

```
[[4 5 6]  
 [7 8 9]]
```

after:

```
[[1 2 3]
 [4 5 6]
 [7 8 9]]
```

**# stack 1d array on top of a 2d array**

```
a = np.array([1,2,3])
b = np.array([[4,5,6],[7,8,9]])
x = np.vstack([a,b])
```

```
[[1 2 3]
 [4 5 6]
 [7 8 9]]
```

**hstack (horizontal stack)**

**hstack** stacks a 1 d array on the top of a 2 d array horizontally.

Before:

```
[[4 5 6]
 [7 8 9]]
```

After:

```
[[1 4 5 6]
 [2 7 8 9]]
```

**# stack 1d array on left side of a 2d array**

```
a = np.array([[1],[2]]) # note 1 d array of 1 d array's
b = np.array([[4,5,6],[7,8,9]])
x = np.hstack([a,b])
```

```
[[1 4 5 6]
 [2 7 8 9]]
```

### Split 1d array into a 2d array

We can split the above 1d array into a 2 d array using the numpy **split** function

```
a = np.array([1, 2]) # [1,2]
a = np.split(a,2) # [[1],[2]]
```

The 2 specifies the index slice to split on [0:2]

We get a 2 dimensional array with values [1] and [2]

```
[array([1]), array([2])]
```

We can now use **hstack** more conveniently where a is **[[1],[2]]** using the split function on [1,2]

```
b = np.array([[4,5,6],[7,8,9]])
x = np.hstack([a,b]) # a = [[1],[2]]
```

```
[[1 4 5 6]
 [2 7 8 9]]
```



## Numpy mathematical operations:

We will use the following 2 dimensional arrays:

```
a = np.array([[1,2],[3,4]])
```

```
b = np.array([[5,6],[7,8]])
```

You may use the mathematical operators  $+$ ,  $-$ ,  $*$ ,  $/$  or the mathematical names add, subtract, multiply or divide.

**add** two numpy arrays

```
c = a + b
```

```
c = np.add(a,b)
```

```
[[6,8][10,12]]
```

Note: If you add a list to a numpy array you get a numpy array.

```
c = a + [[5,6][7,8]]
```

```
[[6,8][10,12]]
```

**subtract** 2 numpy arrays

```
c = a - b
```

```
c = np.subtract(a,b)
```

```
[[ -4, -4][ -4, -4]]
```

**multiply** 2 numpy arrays (just multiplies each element not dot product)

```
c = a * b
```

```
c = np.multiply(a,b)
```

```
[[ 5 12] [21 32]]
```

**divide** 2 numpy arrays

```
c = a / b
```

```
c = np.divide(a,b)
```

```
[[0.2  0.33333333]  
[0.42857143  0.5 ]]
```

**square root** a numpy array

```
c = np.sqrt(a)
```

```
[[1.      1.41421356]
 [1.73205081  2. ]]
```

**matrix multiply** two arrays (dot product)

```
product = a[0] * b[0] + a[1] * b [1]
```

```
c = np.matmul(a,b)
```

```
c = a @ b
```

```
[[19 22][43 50]]
```

Take the **dot product** of two arrays (similar to matrix multiply)

```
dot product = a[0] * b[0] + a[1] * b [1]
```

```
c = a.dot(b)
```

```
c = np.dot(a,b)
```

```
[[19 22][43 50]]
```

**matmul** and **dot** are similar but differ in operation for matrices greater than 2 dimensions

**sum** up all elements in a number 1 dimensional array

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

```
s = np.sum(x)
```

```
15
```

**sum** up all the elements of a 2 dimensional numpy array

```
a = np.array([1,2][3,4])
```

```
s = np.sum(a)
```

```
10
```

Compute **sum** of each **column**

```
a = np.array([[1,2],[3,4]])  
s = np.sum(a, axis=0)
```

```
[4 6]
```

axis = 0 means sum of each column per row along **x axis** (horizontal axis)

```
  1    2  
+ 3    +4  
-----  
  4    6
```

Compute **sum** of each **row**

```
a = np.array([[1,2],[3,4]])  
s = np.sum(a, axis=1)
```

```
[3 7]
```

axis = 1 means sum all columns in a row along **y axis** (vertical axes)

```
1 + 2 = 3    3 + 4 = 7
```

**transpose** a 2 dimensional array `[[1,2],[3,4]]`

```
a = np.array([[1,2],[3,4]])
```

```
c = a.T
```

```
[[1 3] [2 4]]
```

inverse of a matrix `[[1,2],[3,4]]`

```
a = np.array([[1,2],[3,4]])
```

```
c = np.linalg.inv(a)
```

```
[[-2.  1. ] [ 1.5 -0.5]]
```

Note: **linalg** is a sub module of numpy that provides additional mathematical algorithms

add a column of 1's to an array [1,2,3,4,5]

```
x = [1,2,3,4,5]
```

```
y = np.ones(5)
```

```
c = np.c_[ x,y ]
```

```
[[1. 1.] [2. 1.] [3. 1.] [4. 1.] [5. 1.]]
```

### Numpy statistic operations

```
a = np.array([1, 3, 2]) # [ 1 3 2]
```

function	Description	Example using	result
<b>Min</b>	Minimum value of array	<b>a.min()</b>	1
<b>Max</b>	Maximum value of array	<b>a.max()</b>	3
<b>argmin</b>	Index of minimum value	<b>a.argmin()</b>	0
<b>argmax</b>	Index of maximum value	<b>a.argmax()</b>	1
<b>Mean</b>	mean value of array	<b>a.mean()</b>	2
<b>median</b>	median value of array	<b>np.median(a)</b>	2
<b>Std</b>	Standard deviation	<b>a.std()</b>	.82915619758885995

### Median of 2 dimensional array

(median of all elements in array)

```
b = np.array([[1, 2, 3], [5, 6, 1]])  
np.median(b)
```

```
2.5
```

## Median of 2 dimensional array along x axis

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

```
np.median(b, axis=0)
```

```
[3. 4. 2.]
```

median  $(1 + 5)/2 = 3$

median  $(2 + 6) / 2 = 4$

median  $(3 + 1) / 2 = 2$

## Median of 2 dimensional array along y axis

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

```
np.median(b, axis=1)
```

```
[2. 5.]
```

median  $(1 + 2 + 3) = 2$

median  $(5 + 6 + 1) = 5$

## Homework to do

1. Add a 1 dimensional numpy array to a 2 dimensional numpy array. Use **arrange** to make the 1 dimensional array and **linspace** to make the 2 dimensional array.
2. Concatenate two 2 dimensional numpy array together and then take the square root of each element in the result
3. **vstack** a 3 element numpy 1 dimensional array with a 2 by 3 element numpy array then **hstack** the 1 by 3 element 1 dimensional array to the result. Use **split** on the 1 dimensional array.
4. Matrix multiply the inverse of a 3 by 3 element 2numbybarray with the transpose of a 3 x 3 element numpy array
5. Create a 3x3 numpy array with mean 0 and standard deviation 5, then calculate the median using the x axis then the y axis.

Put all your answers in a file called numpyhomework.py

END