

## NUMPY

What are axes?

For (7,) 7 columns

For (2,3) 2 rows and 3 columns

For (2,3,4) 2 depth 3 rows and 4 columns

Note the outermost square brackets refer to the first dimension and so on

If for a thing like `np.sum(axis = 0)` i do then it sums along that axis ie that axis is deleted from the array

So (3,4) becomes (3,) if axis = 1

And (3,4) becomes (4,) if axis = 0

Note if you want it to instead of deleting make the dimension to 1 then use `keepdims = True`

Broadcasting?

Trick to understand broadcasting

$(N,1,2) - (K,2) = (N,K,2)$

It kinda takes lcm of the two corresponding digits basically to give final size.

Imp functions

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

```
b = a.copy() #independent copy
```

```
c = a[1:] #not independent copy so any changes in c will reflect in a
```

You can use `.base` to figure out

`b.base` is `None` while `c.base` is `a`

`np.arange(start, stop, step)` step tells spes between each number

`np.linspace(start, stop, num)` num is number of elements

Slicing:-

start : stop : step for each dimension (:step is optional)

Use just : to denote all elements of a dimension

Can use ... to denote all the other dimensions

Reshaping:-

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

```
flat = arr.flatten() # gives [1,2,3,4] return copy
```

```
rav = arr.ravel() # same but returns view so changes in rav affect arr
```

```
transposed = arr.T
```

```
np.transpose(arr, axes=(1, 0, 2)) # as transposing is just exchanging axes
```

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

```
squeezed = np.squeeze(a) # shape: (3,)
```

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

```
b = np.expand_dims(a, axis=0) # shape: (1, 3)
```

```
b = np.expand_dims(a, axis=1) # shape: (3, 1)
```

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

```
reshaped = arr.reshape(2, 3) #gives [[1 2 3] [4 5 6]]
```

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

```
resized = np.resize(arr, (3, 3))
```

```
# Output:
```

```
# [[1 2 3]
```

```
# [4 5 1]
```

```
# [2 3 4]]
```

Misc

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

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

```
result = np.concatenate((a, b)) # [1 2 3 4 5 6]
```

```
a2 = np.array([[1, 2], [3, 4]])
```

```
b2 = np.array([[5, 6], [7, 8]])
```

```
result_2d = np.concatenate((a2, b2), axis=1)
```

```
# [[1 2 5 6]
```

```
# [3 4 7 8]]
```

The way it works is that it will add up the dimension number along that axis SO if we have (2,a,4) and (2,b,4) and add along axis 1 then we get (2,a+b,4)

**Note:** You cannot concatenate arrays with different shapes unless they are aligned along the specified axis. So remain dimensions have to be same

stack creates a new dimension at a index and stacks along that  
Eg `stacked_col = np.stack((a, b), axis=1)`

Split does even splitting  
`split_result = np.split(a, 3)`  
`print(split_result)`  
`# [array([1, 2]), array([3, 4]), array([5, 6])]`

`Array_split` is similar to original BUT similar to `split()`, but allows **uneven splits** when the array cannot be divided exactly. Remaining elements are distributed as evenly as possible.

WHERE imp  
`np.where(condition, x, y)`  
At position where condition is true you get x and where wrong you get y

`sort()`  
`np.sort(arr, axis)`

Boolean indexing is indexing via a boolean array of same shape of array