

# 02-NumPy-Operations

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## 1 NumPy Operations

### 1.1 Arithmetic

You can easily perform *array with array* arithmetic, or *scalar with array* arithmetic. Let's see some examples:

```
[1]: import numpy as np
      arr = np.arange(0,10)
      arr
```

```
[1]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
[2]: arr + arr
```

```
[2]: array([ 0,  2,  4,  6,  8, 10, 12, 14, 16, 18])
```

```
[3]: arr * arr
```

```
[3]: array([ 0,  1,  4,  9, 16, 25, 36, 49, 64, 81])
```

```
[4]: arr - arr
```

```
[4]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0])
```

```
[5]: # This will raise a Warning on division by zero, but not an error!
      # It just fills the spot with nan
      arr/arr
```

```
C:\Anaconda3\envs\tsa_course\lib\site-packages\ipykernel_launcher.py:3:
RuntimeWarning: invalid value encountered in true_divide
```

This is separate from the ipykernel package so we can avoid doing imports until

```
[5]: array([nan, 1., 1., 1., 1., 1., 1., 1., 1., 1.])
```

```
[6]: # Also a warning (but not an error) relating to infinity  
1/arr
```

C:\Anaconda3\envs\tsa\_course\lib\site-packages\ipykernel\_launcher.py:2:  
RuntimeWarning: divide by zero encountered in true\_divide

```
[6]: array([          inf, 1.          , 0.5          , 0.33333333, 0.25          ,  
          0.2          , 0.16666667, 0.14285714, 0.125          , 0.11111111])
```

```
[7]: arr**3
```

```
[7]: array([ 0,  1,  8, 27, 64, 125, 216, 343, 512, 729], dtype=int32)
```

## 1.2 Universal Array Functions

NumPy comes with many [universal array functions](#), or ufuncs, which are essentially just mathematical operations that can be applied across the array. Let's show some common ones:

```
[8]: # Taking Square Roots  
np.sqrt(arr)
```

```
[8]: array([0.          , 1.          , 1.41421356, 1.73205081, 2.          ,  
          2.23606798, 2.44948974, 2.64575131, 2.82842712, 3.          ])
```

```
[9]: # Calculating exponential (e^)  
np.exp(arr)
```

```
[9]: array([1.00000000e+00, 2.71828183e+00, 7.38905610e+00, 2.00855369e+01,  
          5.45981500e+01, 1.48413159e+02, 4.03428793e+02, 1.09663316e+03,  
          2.98095799e+03, 8.10308393e+03])
```

```
[10]: # Trigonometric Functions like sine  
np.sin(arr)
```

```
[10]: array([ 0.          , 0.84147098, 0.90929743, 0.14112001, -0.7568025 ,  
          -0.95892427, -0.2794155 , 0.6569866 , 0.98935825, 0.41211849])
```

```
[11]: # Taking the Natural Logarithm  
np.log(arr)
```

```
C:\Anaconda3\envs\tsa_course\lib\site-packages\ipykernel_launcher.py:2:
RuntimeWarning: divide by zero encountered in log
```

```
[11]: array([      -inf, 0.          , 0.69314718, 1.09861229, 1.38629436,
          1.60943791, 1.79175947, 1.94591015, 2.07944154, 2.19722458])
```

### 1.3 Summary Statistics on Arrays

NumPy also offers common summary statistics like sum, mean and max. You would call these as methods on an array.

```
[12]: arr = np.arange(0,10)
      arr
```

```
[12]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
[13]: arr.sum()
```

```
[13]: 45
```

```
[14]: arr.mean()
```

```
[14]: 4.5
```

```
[15]: arr.max()
```

```
[15]: 9
```

Other summary statistics include:

### 1.4 Axis Logic

When working with 2-dimensional arrays (matrices) we have to consider rows and columns. This becomes very important when we get to the section on pandas. In array terms, axis 0 (zero) is the vertical axis (rows), and axis 1 is the horizontal axis (columns). These values (0,1) correspond to the order in which `arr.shape` values are returned.

Let's see how this affects our summary statistic calculations from above.

```
[16]: arr_2d = np.array([[1,2,3,4],[5,6,7,8],[9,10,11,12]])
      arr_2d
```

```
[16]: array([[ 1,  2,  3,  4],
          [ 5,  6,  7,  8],
          [ 9, 10, 11, 12]])
```

```
[17]: arr_2d.sum(axis=0)
```

```
[17]: array([15, 18, 21, 24])
```

By passing in `axis=0`, we're returning an array of sums along the vertical axis, essentially  $[(1+5+9), (2+6+10), (3+7+11), (4+8+12)]$

```
[18]: arr_2d.shape
```

```
[18]: (3, 4)
```

This tells us that `arr_2d` has 3 rows and 4 columns.

In `arr_2d.sum(axis=0)` above, the first element in each row was summed, then the second element, and so forth.

So what should `arr_2d.sum(axis=1)` return?

```
[ ]: # THINK ABOUT WHAT THIS WILL RETURN BEFORE RUNNING THE CELL!  
arr_2d.sum(axis=1)
```

## 2 Great Job!

That's all we need to know for now!