```
# create an array of even numbers between 2 and 10
my \ array = np.arange(2,11,2)
'Output': array([ 2, 4, 6, 8, 10])
# sum of array elements
np.sum(my array) # or my array.sum()
'Output': 30
# square root
np.sqrt(my array)
'Output': array([ 1.41421356, 2.
                                      , 2.44948974, 2.82842712,
                 3.16227766])
# log
np.log(my array)
'Output': array([ 0.69314718, 1.38629436, 1.79175947, 2.07944154,
                  2.30258509])
# exponent
np.exp(my array)
'Output': array([ 7.38905610e+00, 5.45981500e+01,
                                                      4.03428793e+02,
                  2.98095799e+03, 2.20264658e+04])
```

Higher-Dimensional Arrays

As we've seen earlier, the strength of NumPy is its ability to construct and manipulate n-dimensional arrays with highly optimized (i.e., vectorized) operations. Previously, we covered the creation of 1-D arrays (or vectors) in NumPy to get a feel of how NumPy works.

This section will now consider working with 2-D and 3-D arrays. 2-D arrays are ideal for storing data for analysis. Structured data is usually represented in a grid of rows and columns. And even when data is not necessarily represented in this format, it is often transformed into a tabular form before doing any data analytics or machine learning. Each column represents a feature or attribute and each row an observation.

Also, other data forms like images are adequately represented using 3-D arrays. A colored image is composed of $n \times n$ pixel intensity values with a color depth of three for the red, green, and blue (RGB) color profiles.

Creating 2-D Arrays (Matrices)

Let us construct a simple 2-D array.

Let's explore common methods in practice for creating 2-D NumPy arrays, which are also matrices.