



# NumPy

A note by  
**Shah Md. Arshad Rahman Ziban**

Declare (or import) system  
 ↓  
**import numpy as np**  
**# Creating a 2D NumPy array**  
 a = np.array([[1, 2, 3], [4, 5, 6]])  
 b = a.view()

Operation	Code	Output	Explanation
Creating a 2D NumPy Array	<code>print(a)</code>	<code># [[1 2 3] # [4 5 6]]</code>	Creates a 2D array with two rows and three columns.
Number of Dimensions	<code>print(a.ndim)</code>	<code># 2</code>	Returns the number of dimensions of the array
Shape of the Array	<code>print(a.shape)</code>	<code># (2,3)</code>	Returns the shape of the array as a tuple (rows, columns).
Size of the Array	<code>print(a.size)</code>	<code># 6</code>	Returns the total number of elements in the array.
Data Type of the Array Elements	<code>print(a.dtype)</code>	<code>#int32 or int64 (may vary based on your system)</code>	Returns the data type of the elements in the array.
Element Size in Bytes	<code>print(a.itemsize)</code>	<code>4 or 8 (may vary based on your system)</code>	Returns the size in bytes of each element in the array.
Bytes Consumed by the Array	<code>print(a.nbytes)</code>	<code>#24 or 48 (may vary)</code>	Returns the total number of bytes consumed by the elements
Transposed Array	<code>print(a.T)</code>	<code># [[1 4] # [2 5] # [3 6]]</code>	Returns the transposed version of the array. Rows become columns and vice versa.
Base Object	<code>print( b.base is a)</code>	<code># True</code>	Checks if the array b shares data with array a (i.e., they share the same base object).
Memory Layout Information	<code>print( a.flags)</code>	<code>#Output will vary</code>	Returns a dictionary-like object with memory layout information of the array.
Strides	<code>print(a.strides)</code>	<code>#Output will vary</code>	Returns a tuple of bytes to step in each dimension when traversing an array.
Real Part	<code>print(a.real)</code>	<code># [[1 2 3] # [4 5 6]]</code>	Returns the real part of the elements in the array (same as original for real numbers).
Imaginary Part	<code>print(a.imag)</code>	<code># [[0 0 0] # [0 0 0]]</code>	Returns the imaginary part of the elements in the array (zero for real numbers).
Flat Iterator	<code>print(list(a.flat))</code>	<code># [1, 2, 3, 4, 5, 6]</code>	Returns a 1D iterator over the elements of the array.
Array as List	<code>print(a.tolist())</code>	<code># [[1, 2, 3], [4, 5, 6]]</code>	Converts the array into a nested list.

Operation	Code	Output	Explanation
Creating an Array of Zeros	<code>zeros_arr = np.zeros((3, 3)) print(zeros_arr)</code>	# [[0. 0. 0.] # [0. 0. 0.] # [0. 0. 0.]]	Creates a 3x3 array filled with zeros.
Creating an Array of Ones	<code>ones_arr = np.ones((2, 4)) print(ones_arr)</code>	# [[1. 1. 1. 1.] # [1. 1. 1. 1.]]	Creates a 2x4 array filled with ones.
Create an array of ones with a specific data type	<code>array_int = np.ones((2, 3), dtype=int) print(array_int)</code>	# [[1 1 1] # [1 1 1]]	Creates a 2x3 array of ones with ones with a specific data type.
Creating an Empty Array (Values are Not Initialized)	<code>empty_arr = np.empty((2, 3)) print(empty_arr)</code>	# Output varies	Creates a 2x3 array with uninitialized values (values are whatever is already in memory).
Creating an Array (1D) with a Range of Values	<code>arange_arr = np.arange(0, 10, 2) print(arange_arr)</code>	# [0 2 4 6 8]	Creates a 1D array with values from 0 to 8, with a step of 2.
Creating an Array (2D) with a Range of Values	<code>array_2d = np.arange(1, 13).reshape(3, 4) print(array_2d)</code>	# [[1 2 3 4] # [5 6 7 8] # [9 10 11 12]]	<code>np.arange(1, 13)</code> generates numbers from 1 to 12, and <code>.reshape(3, 4)</code> formats them into a 3x4 array.
Creating an Array with Evenly Spaced Values	<code>linspace_arr = np.linspace(0, 1, 5) print(linspace_arr)</code>	# [0. 0.25 0.5 0.75 1.]	Creates a 1D array with 5 evenly spaced values between 0 and 1.
Creating an Array with Random Values	<code>random_arr = np.random.random((2, 3)) print(random_arr)</code>	# Output will vary due to random values	Creates a 2x3 array with random values between 0 and 1.
Creating an Array with Random Integers	<code>random_int_arr = np.random.randint(0, 10, (3, 3)) print(random_int_arr)</code>	# Output will vary due to random values	Creates a 3x3 array with random integers between 0 and 9.
Creating an Identity Matrix	<code>identity_matrix = np.eye(3) print(identity_matrix)</code>	# [[1. 0. 0.] # [0. 1. 0.] # [0. 0. 1.]]	Creates a 3x3 identity matrix (1s on the diagonal, 0s elsewhere).

## Indexing

Example Type	Code
1D Array	<pre>arr = np.array([1, 2, 3, 4, 5]) print(arr[0]) # Output: 1 print(arr[-1]) # Output: 5</pre>
2D Array	<pre>arr_2d = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]]) print(arr_2d[1, 2]) # Output: 6 print(arr_2d[-2, -1]) # Output: 6</pre>
Boolean	<pre>arr = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]]) # Creating a boolean array bool_idx = (arr &gt; 5) print("Boolean array:\n", bool_idx) # Output: # Boolean array: # [[False False False] #  [False False True] #  [ True  True True]]</pre> <pre># Using the boolean array to index the original array filtered_arr = arr[bool_idx] print("Elements greater than 5:", filtered_arr) # Output: Elements greater than 5: [6 7 8 9]</pre>
Fancy	<pre>arr = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]]) # Using integer arrays to index rows = np.array([0, 1, 2]) cols = np.array([2, 1, 0]) fancy_indexed_arr = arr[rows, cols] print("Fancy indexed array:", fancy_indexed_arr) # Output: Fancy indexed array: [3 5 7]</pre>

## slicing/subsetting

Example Type	Code
1D Array	<pre>arr = np.array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9]) sliced = arr[2:6] print(sliced) # Output: [2 3 4 5]</pre>
2D Array	<pre>arr = np.array([[0, 1, 2], [3, 4, 5], [6, 7, 8]]) sliced = arr[0:2, 0:2] print(sliced) # Output: # [[0 1] # [3 4]]</pre>
Using Step (1D)	<pre>arr = np.array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9]) sliced = arr[1:9:2] print(sliced) # Output: [1 3 5 7]</pre>
Using Step (2D)	<pre>arr = np.array([[1, 2, 3, 4],[5, 6, 7, 8],[9, 10, 11, 12]]) sub_array = arr[1, ::2] print(sub_array) # Output: [5 7]</pre>
Negative Indexing	<pre>arr = np.array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9]) sliced = arr[-5:] print(sliced) # Output: [5 6 7 8 9]</pre>
Modifying a Slice	<pre>arr = np.array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9]) arr[2:6] = [20, 30, 40, 50] print(arr) #Output: [ 0 1 20 30 40 50 6 7 8 9]</pre>
Slicing with Negative Step	<pre>arr = np.array([[1, 2, 3],[5, 6, 7, 8],[9, 10, 11, 12]]) sub_array = arr[1, ::-1] print(sub_array) # Output: [8 7 6 5]</pre>
Modifying Slices	<pre>arr = np.array([[1, 2, 3],[4, 5, 6],[7, 8, 9]]) arr[:2, 1:3] = 99 print( arr) # Output: # [[1 99 99] # [4 99 99] # [7 8 9]]</pre>

Operation	Code	Description
Reshape	<pre>arr1 = np.array([1, 2, 3, 4, 5, 6]) reshaped_arr = arr1.reshape((2, 3)) print(reshaped_arr) #Output: #[[1 2 3] #[4 5 6]]</pre>	Reshapes the array without modifying the original array.
Resize (in-place)	<pre>arr1 = np.array([1, 2, 3, 4, 5, 6]) arr1.resize((2, 3)) print(arr1) # Output: #[[1 2 3] #[4 5 6]]</pre>	Changes the shape of the array in-place.
Resize (larger shape)	<pre>arr1 = np.array([1, 2, 3, 4, 5, 6]) # refcheck=False allows resizing even if the array #is referenced elsewhere arr1.resize((3, 3), refcheck=False) print(arr1) # Output may vary # Example output: #[[1 2 3] #[4 5 6] #[1 2 3]]</pre>	Resizes the array to a larger shape, potentially filling with repeated data.
Flatten	<pre>arr1 = np.array([[1, 2, 3], [4, 5, 6]]) flattened_arr = arr1.flatten() print(flattened_arr) # Output: #[1 2 3 4 5 6]</pre>	Returns a new 1D array that is a copy of the original array's elements.
Ravel	<pre>arr1 = np.array([[1, 2, 3], [4, 5, 6]]) raveled_arr = arr1.ravel() print(raveled_arr) # Output: #[1 2 3 4 5 6]</pre>	Returns a 1D array that is a view of the original array whenever possible. Modifications to the raveled array can affect the original array if a view is returned.

Operation	Code	Description	
Horizontal Stack (hstack)	<pre>import numpy as np a = np.array([1, 2, 3]) b = np.array([4, 5, 6]) c = np.hstack((a, b)) print(c)</pre> <p># Output: [1 2 3 4 5 6]</p>	Stacks arrays in sequence horizontally (column-wise).	
Vertical Stack (vstack)	<pre>import numpy as np a = np.array([1, 2, 3]) b = np.array([4, 5, 6]) c = np.vstack((a, b)) print(c)</pre> <p># Output: [[1 2 3] [4 5 6]]</p>	Stacks arrays in sequence vertically (row-wise).	
Column Stack	<pre>import numpy as np a = np.array([1, 2, 3]) b = np.array([4, 5, 6]) c = np.column_stack((a, b)) print(c)</pre> <p># Output: [[1 4] [2 5] [3 6]]</p>	Stacks 1-D arrays as columns into a 2-D array. For 2-D arrays, behaves like hstack.	
Row Stack	<pre>import numpy as np a = np.array([1, 2, 3]) b = np.array([4, 5, 6]) c = np.row_stack((a, b)) print(c)</pre> <p># Output: [[1 2 3] [4 5 6]]</p>	Equivalent to vstack for 1-D arrays, stacking them along a new row axis.	
Depth Stack (dstack)	<pre>import numpy as np a = np.array([[1, 2, 3], [4, 5, 6]]) b = np.array([[7, 8, 9], [10, 11, 12]]) c = np.dstack((a, b)) print(c)</pre> <p># Output: [[[1 7] [2 8] [3 9]] [[4 10] [5 11] [6 12]]]</p>	Stacks arrays in sequence depth-wise (along the third dimension).	
General Stack (stack)	<pre>import numpy as np a = np.array([1, 2, 3]) b = np.array([4, 5, 6]) c = np.stack((a, b), axis=0) print(c) # Output: [[1 2 3] [4 5 6]]</pre> <pre>c = np.stack((a, b), axis=1) print(c) # Output: [[[1 4] [2 5] [3 6]]]</pre>	<pre>import numpy as np a = np.array([[1, 2, 3], [4, 5, 6]]) b = np.array([[7, 8, 9], [10, 11, 12]]) c = np.stack((a, b), axis=2) print(c) # Output: [[[1 7] [2 8] [3 9]] [[4 10] [5 11] [6 12]]]</pre>	Joins a sequence of arrays along a new axis. The axis parameter specifies the index of the new axis.

## Adding

Operation	Code	Description
Appending Elements	<pre>import numpy as np  arr = np.array([1, 2, 3]) values_to_append = [4, 5, 6] appended_arr = np.append(arr, values_to_append)  print(appended_arr) # Output: [1 2 3 4 5 6]</pre>	Appends values to the end of an array. Does not modify the original array but returns a new array with the appended values. Can be used with 1D and multi-dimensional arrays.
	<pre>import numpy as np  arr = np.array([[1, 2, 3], [4, 5, 6]]) values_to_append = np.array([[7, 8, 9], [10, 11, 12]])  appended_arr = np.append(arr, values_to_append, axis=0) print(appended_arr) # Output: # [[1 2 3] # [4 5 6] # [7 8 9] # [10 11 12]]  values_to_append = np.array([[13], [14]]) appended_arr = np.append(arr, values_to_append, axis=1) print(appended_arr) # Output: # [[1 2 3 13] # [4 5 6 14]]</pre>	
Inserting Elements	<pre>import numpy as np  arr = np.array([1, 2, 3, 4, 5]) values_to_insert = [9, 8, 7] inserted_arr = np.insert(arr, 2, values_to_insert)  print(inserted_arr) # Output: [1 2 9 8 7 3 4 5]</pre>	Inserts values into an array before the specified index. Does not modify the original array but returns a new array with the inserted values. Can be used with 1D and multi-dimensional arrays. The index specifies the position in the array where the new values should be inserted.
	<pre>import numpy as np  arr = np.array([[1, 2, 3], [4, 5, 6]]) values_to_insert = np.array([[7, 8, 9]])  # Inserting values into the array before index 1 along axis 0 inserted_arr = np.insert(arr, 1, values_to_insert, axis=0) print(inserted_arr) # Output: # [[1 2 3] # [7 8 9] # [4 5 6]]  # Inserting values into the array before index 2 along axis 1 values_to_insert = np.array([[10], [11]]) inserted_arr = np.insert(arr, 2, values_to_insert, axis=1) print(inserted_arr) # Output: # [[1 2 10 3] # [4 5 11 6]]</pre>	

## Removing

Operation	Code	Description
Deleting Rows	<pre>import numpy as np arr_2d = np.array([[1, 2, 3],                   [4, 5, 6],                   [7, 8, 9]]) deleted_element_arr_2d = np.delete(arr_2d, 4) print(deleted_element_arr_2d)  # Output: # [1 2 3 4 6 7 8 9]  arr_1d = np.array([1, 2, 3, 4, 5, 6, 7, 8]) deleted_elements_arr_1d = np.delete(arr_1d, [2, 4]) print(deleted_elements_arr_1d)  # Output: # [1 2 4 5 7 8]</pre>	<p>In the 2D array example, a specific element was deleted after flattening the array.</p> <p>In the 1D array example, specific elements were removed based on their indices.</p>
	<pre>import numpy as np arr_2d = np.array([[1, 2, 3],                   [4, 5, 6],                   [7, 8, 9],                   [10, 11, 12]])  # Deleting the row at index 3 deleted_row_arr = np.delete(arr_2d, 3, axis=0) print(deleted_row_arr)  # Output: # [[1 2 3] # [4 5 6] # [7 8 9]]</pre>	<p>The original array had 4 rows. The row at index 3 (4th row) [10, 11, 12] was deleted. The resulting array now has 3 rows.</p>
Deleting Columns	<pre>import numpy as np  arr_2d = np.array([[1, 2, 3, 4],                   [5, 6, 7, 8],                   [9, 10, 11, 12]])  # Deleting the column at index 3 (4th column) deleted_col_arr = np.delete(arr_2d, 3, axis=1) print(deleted_col_arr)  # Output: # [[1 2 3] # [5 6 7] # [9 10 11]]</pre>	<p>This deletes the column at index 3 in the original array. The result is an array with the last column removed.</p>

Operation	Code	Description
Loading from a text file	<pre>import numpy as np  # Loading data from a text file data = np.loadtxt('file.txt')  # Printing the loaded data print(data)</pre>	Loads data from a text file (file.txt) and prints it. Assumes the file contains space-separated values.
Loading from a CSV file	<pre>import numpy as np  # Loading data from a CSV file data = np.genfromtxt('file.csv', delimiter=',')  # Printing the loaded data print(data)</pre>	Loads data from a CSV file (file.csv) using a comma as a delimiter and prints it.  The delimiter=',' parameter specifies that the values in the file are separated by commas.
Writing to a text file	<pre>import numpy as np  # Data to be written to the text file arr = np.array([[1, 2, 3], [4, 5, 6]])  np.savetxt('file.txt', arr, delimiter=' ')</pre>	Writes a NumPy array to a text file (file.txt) with space as the delimiter.
Writing to a CSV file	<pre>import numpy as np  # Data to be written to the CSV file arr = np.array([[1, 2, 3], [4, 5, 6]])  np.savetxt('file.csv', arr, delimiter=',')</pre>	Writes a NumPy array to a CSV file (file.csv) with a comma as the delimiter.

Operation	Code	Description
Concatenating Arrays Along Rows	<pre>import numpy as np  arr1 = np.array([[1, 2], [3, 4]]) arr2 = np.array([[5, 6], [7, 8]]) result = np.concatenate((arr1, arr2), axis=0) print(result)  # Output: # [[1 2] #  [3 4] #  [5 6] #  [7 8]]</pre>	Combines arrays along rows (vertically)
Concatenating Arrays Along Columns	<pre>import numpy as np  arr1 = np.array([[1, 2], [3, 4]]) arr2 = np.array([[5, 6], [7, 8]])  result = np.concatenate((arr1, arr2), axis=1) print(result)  # Output: # [[1 2 5 6] #  [3 4 7 8]]</pre>	Combines arrays along columns (horizontally)
Splitting Array into Sub-arrays	<pre>import numpy as np  arr = np.array([1, 2, 3, 4, 5, 6, 7, 8, 9])  result = np.split(arr, 3) print(result)  # Output: # [array([1, 2, 3]), array([4, 5, 6]), array([7, 8, 9])]</pre>	Splits a 1D array into 3 equal-sized sub-arrays
Splitting Horizontally at Index	<pre>import numpy as np arr = np.array([[1, 2, 3, 4, 5],                [6, 7, 8, 9, 10]])  result = np.hsplit(arr, 5) print(result)  # Output: # [array([[1], #         [6]]), #  array([[2], #         [7]]), #  array([[3], #         [8]]), #  array([[4], #         [9]]), #  array([[5], #         [10]])]</pre>	Splits a 2D array horizontally into sub-arrays at columns

Operation	Code	Description
Elementwise Addition	<pre>arr1 = np.array([4, 5, 6]) arr2 = np.array([1, 2, 3])  result = np.add(arr1, arr2) print(result) # Output: [5 7 9]</pre>	Adds elements of two arrays elementwise
Elementwise Subtraction	<pre>result = np.subtract(arr1, arr2) print(result) # Output: [3 3 3]</pre>	Subtracts elements of two arrays elementwise
Elementwise Multiplication	<pre>result = np.multiply(arr1, arr2) print(result) # Output: [ 4 10 18]</pre>	Multiplies elements of two arrays elementwise
Elementwise Division	<pre>result = np.divide(arr1, arr2) print(result) # Output: [4. 2.5 2.]</pre>	Divides elements of two arrays elementwise
Elementwise Power	<pre>result = np.power(arr1, arr2) print(result) # Output: [ 4 25 216]</pre>	Raises elements of one array to powers of another
Square Root of Elements	<pre>arr = np.array([1, 4, 9, 16]) result = np.sqrt(arr) print(result) # Output: [1. 2. 3. 4.]</pre>	Finds square root of each element
Sine of Elements	<pre>arr = np.array([0, np.pi / 2, np.pi]) result = np.sin(arr) print(result) # Output: [0. 1. 0.]</pre> <p>0, <math>\pi/2</math> (90 degrees), and <math>\pi</math> (180 degrees).</p>	Computes sine of each element
Natural Log of Elements	<pre>arr = np.array([1, np.e, np.e**2]) result = np.log(arr) print(result) # Output: [0. 1. 2.]</pre> <p><math>\log(1)</math> is 0.  <math>\log(e)</math> is 1  <math>\log(e^2)</math> is 2</p>	Computes natural log of each element
Absolute Value of Elements	<pre>arr = np.array([-1, -2, 3, -4]) result = np.abs(arr) print(result) # Output: [1 2 3 4]</pre>	Computes absolute value of each element
Round Up to Nearest Integer	<pre>arr = np.array([1.1, 2.5, 3.7]) result = np.ceil(arr) print(result) # Output: [2. 3. 4.]</pre>	Rounds up each element to nearest integer
Round Down to Nearest Integer	<pre>result = np.floor(arr) print(result) # Output: [1. 2. 3.]</pre>	Rounds down each element to nearest integer
Round to Nearest Integer	<pre>arr = np.array([1.1, 2.5, 3.7]) result = np.round(arr) print(result) # Output: [1. 2. 4.]</pre> <p>1.1 is closer to 1 than to 2, so it rounds to 1. When a number is exactly halfway between two integers (e.g., 2.5), it rounds to the nearest even integer. This is known as "round half to even".</p>	Rounds each element to nearest integer

Operation	Code	Description
Add a scalar to each element	<pre>arr = np.array([1, 2, 3]) result = np.add(arr, 1) print(result) # Output: [2 3 4]</pre>	Adds a scalar to each element of the array
Subtract a scalar from each element	<pre>arr = np.array([4, 5, 6]) result = np.subtract(arr, 2) print(result) # Output: [2 3 4]</pre>	Subtracts a scalar from each element of the array
Multiply each element by a scalar	<pre>arr = np.array([1, 2, 3]) result = np.multiply(arr, 3) print(result) # Output: [3 6 9]</pre>	Multiplies each element by a scalar
Divide each element by a scalar	<pre>arr = np.array([4, 5, 6]) result = np.divide(arr, 4) print(result) # Output: [1. 1.25 1.5 ]</pre>	Divides each element by a scalar
Raise each element to the specified power	<pre>arr = np.array([1, 2, 3]) result = np.power(arr, 5) print(result) # Output: [ 1 32 243]</pre>	Raises each element to the specified power

Operation	Code	Description
Mean along specific axis	<pre>arr = np.array([[1, 2, 3], [4, 5, 6]]) mean = np.mean(arr, axis=0) print(mean) # Output: [2.5 3.5 4.5]</pre>	Computes the mean along the specified axis (0 for columns)
Sum of elements	<pre>arr = np.array([1, 2, 3, 4]) total = arr.sum() print(total) # Output: 10</pre>	Computes the sum of all elements in the array
Minimum value	<pre>arr = np.array([1, 2, 3, 4]) min_val = arr.min() print(min_val) # Output: 1</pre>	Finds the minimum value in the array
Maximum value along specific axis	<pre>arr = np.array([[1, 2, 3], [4, 5, 6]]) max_val = arr.max(axis=0) print(max_val) # Output: [4 5 6]</pre>	Computes the maximum value along the specified axis (0 for columns)
Variance of array	<pre>arr = np.array([1, 2, 3, 4]) variance = np.var(arr) print(variance) # Output: 1.25</pre>	Computes the variance of the array
Standard deviation along specific axis	<pre>arr = np.array([[1, 2, 3], [4, 5, 6]]) std_dev = np.std(arr, axis=1) print(std_dev) # Output: [0.81649658 0.81649658]</pre>	Computes the standard deviation along the specified axis (1 for rows)
Correlation coefficient of array	<pre>arr = np.array([[1, 2, 3], [4, 5, 6]]) corr_coef = np.corrcoef(arr) print(corr_coef) # Output: # [[1. 1.] # [1. 1.]]</pre>	Computes the correlation coefficient matrix of the array