

Chapter 4 - NumPy Basics: Arrays and Vectorized Computation

W. Mckinney, "Numpy Basics: Arrays and Vectorized Computation," in *Python for Data Analysis: Data Wrangling with Pandas, NumPy, and Jupyther*, 3rd ed., O'Reilly, 2022, pp. 83-121.

https://datamineaz.org/



Numpy

- NumPy, short for Numerical Python, is one of the most important foundational packages for numerical computing in Python.
 - ndarray, an efficient multidimensional array providing fast array-oriented arithmetic operations and flexible broadcasting capabilities
 - Mathematical functions for fast operations on entire arrays of data without having to write loops
 - Tools for reading/writing array data to disk and working with memory-mapped files
 - Linear algebra, random number generation, and Fourier transform capabilities
 - A C API for connecting NumPy with libraries written in C, C++, or FORTRAN

Numpy Arrays

Creating arrays

```
# Range of values
range_arr = np.arange(10)
range_arr
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
# Array of zeros
zeros arr = np.zeros((4, 3))
zeros arr
array([[0., 0., 0.],
       [0., 0., 0.],
       [0., 0., 0.],
       [0., 0., 0.]
# Array of ones
ones arr = np.ones((3, 2))
ones arr
array([[1., 1.],
       [1., 1.],
       [1., 1.]])
 # Identity matrix
 identity matrix = np.eye(3)
 identity matrix
 array([[1., 0., 0.],
        [0., 1., 0.],
        [0., 0., 1.]])
```

Numpy Arrays

Array attributes

```
# Multidimensional array
multi arr = np.array([[1, 2, 3], [4, 5, 6]])
# Array dimensions
print("Dimensions:", multi arr.ndim)
# Shape of array
print("Shape:", multi_arr.shape)
# Size of array
print("Size:", multi_arr.size)
# Data type of array elements
print("Data Type:", multi arr.dtype)
```

Dimensions: 2

Shape: (2, 3)

Size: 6

Data Type: int64

Numpy Operations

Arithmetic operations

```
import numpy as np
# Creating a simple NumPy array
arr = np.array([1, 2, 3, 4])
# Flement-wise addition
addition = arr + 2
addition
array([3, 4, 5, 6])
# Flement-wise subtraction
subtraction = arr - 2
subtraction
array([-1, 0, 1, 2])
```

```
# Element-wise multiplication
multiplication = arr * 2
multiplication
array([2, 4, 6, 8])
# Element-wise division
division = arr / 2
division
array([0.5, 1. , 1.5, 2. ])
```

Numpy Operations

• Statistical operations

```
import numpy as np

# Creating a simple NumPy array
arr = np.array([1, 2, 3, 4])
# Sum of elements
total = arr.sum()
total
```

np.int64(10)

np.float64(2.5)

```
# Mean of elements
mean_value = arr.mean()
mean_value
```

```
# Standard deviation
std dev = arr.std()
std dev
np.float64(1.118033988749895)
# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])
# Correlation coefficient
corr = np.corrcoef(multi arr)
corr
array([[1., 1.],
       [1., 1.]])
```



Advanced Operations

Reshaping and transposing

```
# Range of values
range_arr = np.arange(10)
print('range_arr: ', range_arr)

# Reshaping an array
reshaped = np.reshape(range_arr, (2, 5))
print('rreshaped: ', reshaped)

range_arr: [0 1 2 3 4 5 6 7 8 9]
rreshaped: [[0 1 2 3 4]
[5 6 7 8 9]]
```

```
# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])
print('multi arr: ', multi arr)
# Transpose of an array
transposed = multi arr.T
print('transposed: ', transposed)
multi arr: [[1 2 3]
 [4 5 6]]
transposed: [[1 4]
 [2 5]
 [3 6]]
```

Advanced Operations

Indexing and slicing

```
# Multidimensional array
multi arr = np.array([[1, 2, 3], [4, 5, 6]])
print('multi_arr: ', multi_arr)
# Accessing a specific element
element = multi arr[0, 1]
element
multi arr: [[1 2 3]
 [4 5 6]]
np.int64(2)
```

```
# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])
print('multi arr: ', multi arr)
# Slicing a row
row = multi arr[1, :]
print('row: ', row)
multi arr: [[1 2 3]
 [4 5 6]]
row: [4 5 6]
# Multidimensional array
multi arr = np.array([[1, 2, 3], [4, 5, 6]])
print('multi arr: ', multi arr)
# Slicing a column
column = multi arr[:, 2]
print('column: ', column)
multi arr: [[1 2 3]
 [4 5 6]]
column: [3 6]
```



Advanced Operations

Broadcasting

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

# Broadcasting allows arithmetic operations on arrays of different sizes
broadcasted_addition = multi_arr + np.array([1, 0, 1])
print('broadcasted_addition: ', broadcasted_addition)
```



Linear Algebra

Matrix operations

```
# Creating a simple NumPy array
arr = np.array([1, 2, 3, 4])

# Dot product
dot_product = np.dot(arr, arr)
print('dot_product: ', dot_product)

# Matrix multiplication
matrix_mul = np.dot(multi_arr, identity_matrix)
print('matrix_mul: ', matrix_mul)

matrix_mul: [[1, 2, 3], [4, 5, 6]])

# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])

# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])

# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])

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# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])

# Multidimensional array
multi_arr = np.array([[1, 2, 3], [4, 5, 6]])

# Matrix_multiplication
matrix_mul: ', matrix_mul)
```



Linear Algebra

• Eigenvalues and Eigenvectors

```
# Identity matrix
identity_matrix = np.eye(3)

# Eigenvalues and eigenvectors
eigenvalues, eigenvectors = np.linalg.eig(identity_matrix)

print('eigenvalues: ', eigenvalues)
print('eigenvectors: ', eigenvectors)

eigenvalues: [1. 1. 1.]
eigenvectors: [[1. 0. 0.]
[0. 1. 0.]
[0. 0. 1.]]
```