## numpy-dsp

## August 31, 2024

```
[1]: import numpy as np
[2]: # Creating a 1D array
     arr1 = np.array([1, 2, 3, 4, 5])
     # Creating a 2D array
     arr2 = np.array([[1, 2, 3], [4, 5, 6]])
    basic operation
[3]: # Addition
     arr_sum = arr1 + 10
     print(arr_sum,'\n')
     # Multiplication
     arr_mult = arr1 * 2
     print(arr_mult,'\n')
     # Dot product (for 2D arrays)
     arr_dot = np.dot(arr2, arr1[:3])
     print(arr_dot,'\n')
    [11 12 13 14 15]
    [2 4 6 8 10]
    [14 32]
    Array Properties
[4]: print(arr1.shape, '\n')
     print(arr2.shape,'\n')
     print(arr1.size,'\n')
     print(arr2.dtype,'\n')
    (5,)
    (2, 3)
    5
```

int64

2.Data Manipulation:

```
[5]: # Create a 3x3 array
      arr = np.arange(9).reshape(3, 3)
      print(arr,'\n')
     [[0 1 2]
      [3 4 5]
      [6 7 8]]
 [6]: # Indexing
      element = arr[1, 2] # Get element from 2nd row, 3rd column
      print(element,'\n')
     5
 [7]: # Slicing
      slice_arr = arr[:2, 1:] # First two rows and last two columns
      print(slice_arr,'\n')
     [[1 2]
      [4 5]]
 [8]: # Reshaping
      reshaped_arr = arr.reshape(1, 9) # Reshape to 1x9 array
      print(reshaped_arr,'\n')
     [[0 1 2 3 4 5 6 7 8]]
 [9]: # Element-wise square root
      sqrt_arr = np.sqrt(arr)
      print(sqrt_arr,'\n')
     [[0.
                  1.
                             1.41421356]
      [1.73205081 2.
                             2.23606798]
      [2.44948974 2.64575131 2.82842712]]
[10]: # Logarithm
      log_arr = np.log(arr + 1) # Adding 1 to avoid log(0)
      print(log_arr,'\n')
```

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[1.38629436 1.60943791 1.79175947]
      [1.94591015 2.07944154 2.19722458]]
     3.Data Aggregation
[11]: # Compute mean
      mean_value = np.mean(arr)
      print(mean_value,'\n')
     4.0
[12]: # Compute median
      median_value = np.median(arr)
      print(median_value,'\n')
     4.0
[13]: # Compute standard deviation
      std_dev = np.std(arr)
      print(std_dev,'\n')
     2.581988897471611
[14]: # Grouping: Separate even and odd numbers
      even_numbers = arr[arr % 2 == 0]
      print(even_numbers,'\n')
      odd_numbers = arr[arr % 2 != 0]
      print(odd_numbers,'\n')
     [0 2 4 6 8]
     [1 3 5 7]
[15]: # Aggregation on groups
      sum_even = np.sum(even_numbers)
      print(sum_even,'\n')
      sum_odd = np.sum(odd_numbers)
      print(sum_odd,'\n')
     20
     16
```

[[0.

0.69314718 1.09861229]

4.Data Analysis

```
[17]: #Create two arrays
      x = np.array([1, 2, 3, 4, 5])
      y = np.array([2, 4, 6, 8, 10])
[18]: # Correlation coefficient
      correlation = np.corrcoef(x, y)
      print(correlation,'\n')
     [[1. 1.]
      [1. 1.]]
[19]: data = np.array([1, 2, 3, 4, 100, 6, 7, 8, 9])
      # Calculate the z-scores
      z_scores = (data - np.mean(data)) / np.std(data)
      print(z_scores,'\n')
     [-0.4857185 -0.45234853 -0.41897856 -0.38560858 2.81790887 -0.31886864
      -0.28549866 -0.25212869 -0.21875871]
[20]: #Identify outliers
      outliers = data[np.abs(z_scores) > 2]
      print(outliers,'\n')
     [100]
[21]: # Calculate the 25th, 50th, and 75th percentiles
      percentiles = np.percentile(data, [25, 50, 75])
      print(percentiles,'\n')
     [3. 6. 8.]
     5. Application in Data Science
     Advantages of Using NumPy:
```

Efficiency

:NumPy is faster than regular Python lists due to its underlying C implementation and efficient memory usage.

Convenience: Provides a broad range of mathematical functions.

Interoperability: Integrates seamlessly with other Python libraries like Pandas, Scikit-learn, and TensorFlow, which are heavily used in data science.

Real-World Examples: Machine Learning: NumPy is the backbone of many machine learning libraries. It's used for data preprocessing, implementing algorithms, and even deep learning (e.g., TensorFlow).

Financial Analysis: NumPy's efficient handling of large datasets is critical for analyzing stock prices, calculating moving averages, and performing Monte Carlo simulations.

Scientific Research: Researchers rely on NumPy for processing large datasets, performing simulations, and analyzing experimental data.

Conclusion: NumPy is a powerful tool for any data science professional. Its ability to handle large datasets efficiently and perform complex numerical operations makes it an essential library in the data science toolkit.

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