

# NumPy Arrays

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1405

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# Required Python Libraries for Today

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## Core

- NumPy, Pandas

```
import numpy as np  
import pandas as pd
```

## Visualization

- Matplotlib, Seaborn

```
from matplotlib import pyplot as plt  
import seaborn as sns
```

## Statistical learning

- scikit-learn, SciPy, statsmodels

```
import sklearn  
import scipy  
import statsmodel as sm
```

# NumPy & Pandas Printing Format

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It's better to print only *the first few decimal digits* of large real numbers. Set the print options to keep 3 decimal digits and suppress scientific notation (for NumPy arrays and Pandas DataFrames):

- **NumPy**

```
np.set_printoptions(precision=3, suppress=True)
```

- **Pandas**

```
pd.options.display.float_format = '{:.3f}'.format
```

# NumPy Array

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- A NumPy array is a very useful data structure for storing *large, multi-dimensional, homogenous* data for efficient *numerical computations* (requires importing the `numpy` library)

Key features:

- Homogeneity: (1) values are of the same scalar type and often numeric  
(2) rectangular so all elements have the exact same size
- Multidimensional: 1D, 2D, or higher-dimension (1D & 1D most common)
- Efficiency: faster than lists for numeric computations, and supports built-in mathematic and statistics functions

# NumPy Array Examples

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`['abc',12]` not a NumPy array → 'abc' and 12 are different data types

`[[1,2],[3]]` not a NumPy array → [1,2] and [3] have different sizes

`[[[1,2],[3,4]],[[5,6],[7,8]]]` is a NumPy array of shape (2, 2, 2)

`[0.1, 0.5, 1.0, 1.5]` is a NumPy array of shape (4,) of float type

# 2D Array As a Sequence of 1D Arrays

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- We can think of a 2D array `arr` as either
  - A sequence of 1D arrays as rows (of the same length), or
  - A sequence of 1D arrays as columns (of the same length)
- For `arr[i, j]`, some important concepts and their relationships
  - `i` is a *row* index: `axis=0`, and number of rows equals `arr.shape[0]`
  - `j` is a *column* index: `axis=1` and number of columns equals `arr.shape[1]`

	Rows ( <code>axis=0</code> )	Columns ( <code>axis=1</code> )
Role	Represents an Observation	Represents a Variable
Access	<code>array[i, :]</code> (row $i$ )	<code>array[:, j]</code> (column $j$ )
Size	<code>array.shape[0]</code>	<code>array.shape[1]</code>

# Common Ways to Initialize an Array

- Directly from a list: `np.array([0.1,0.2,0.3,0.4], dtype=np.float64)`

- Zeros: `np.zeros(5, dtype=np.float64)` → `array([0., 0., 0., 0., 0.])`

- Ones: `np.ones(5, dtype=np.float64)` → `array([1., 1., 1., 1., 1.])`

- Uninitialized values (empties)

- `np.empty([2,2], dtype=np.int64)` → An arbitrary 2x2 ndarray

- Sequence with specified *start*, *stop*, & *step*

- `np.arange(10)`

- `np.arange(0.1, 1.05, 0.05)`

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

 `array([0.1 , 0.15, 0.2 , 0.25, 0.3 , 0.35, 0.4 , 0.45, 0.5 ])`

# Common NumPy Data Types (Numeric)

Data type	Type description	Range (int/bool) or Precision (float)	
np.int8	8-bit integer	-128 to 127	
np.int16	16-bit integer	-32768 to 32767	
np.int32	32-bit integer	- $2^{31}$ to $2^{31}-1$ (~9 digits)	
np.int64	64-bit integer	- $2^{63}$ to $2^{63}-1$ (~19 digits)	
np.float16	Half-precision floating point	3-4 decimal digits	
np.float32	Single-precision floating point	7 decimal digits	
np.float64	double-precision floating point	15-16 decimal digits	
np.float128	Extended-precision floating point	33 decimal digits	
np.bool_	Boolean type	{True, False}	Note trailing underscores!
np.string_	Fixed-length ASCII string type		

# NumPy Data Type Casting

- Automatically conversion of array elements of different types:  
bool\_ → int\* → float\* → string\_

For example:

`np.array([1, 'two', 3.0])` → `array(['1', 'two', '3.0'], dtype='<U32')`

Unicode  
string of  
length 32

`np.array([100, 10.0, False])` → `np.array([1,'two',3.0])`

# NumPy Data Type Casting

- Explicit type casting, two approaches `arr=np.array([0, 1, 2, 3, 4])`:
  - `arr.astype(np.float64)`
  - `np.float(arr)`

→ `array([0., 1., 2., 3., 4.])`

*Cast the array into float type and give the same result*
- Caution: casting from coarse to precise is okay (e.g., int to float), but try to avoid casting from precise to coarse, for example
  - Allowed but loses precision: `np.int64(100.9)` *Keeps integer part only*
  - Allowed but messing up completely: `np.int16(1000000)` *Out of range*
  - Not allowed and returns an error: `np.float32('hello world')`

# Representing Missing Values

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A numeric missing value is represented by `np.nan` (“not a number”)

- Note: This does not apply to strings, where missing values are simply denoted by `NoneType` (`None`) or empty string (`""`)
- *Comparison logics don't work* on the missing value
  - Any comparison between a non-missing value `x` and `np.nan` results in `False`: e.g., `np.nan<0` and `np.nan>=0` both evaluate to `False`
- In general, final data *should not contain* `np.nan`; clean the data to remove or fill missing values before proceeding with data analysis

# Exercise: Create Arrays & Cast Types

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1. Create a NumPy array of integers from 0 to 999 with `dtype=np.int16`, then convert it to `dtype=np.float32`
2. Create a NumPy array of 1000 zeros with `dtype=np.float16`, then convert it to `dtype=np.bool`
3. Generate a 2D NumPy array with 100 rows and 3 columns:
  - *Column 0*: row numbers 0-99
  - *Column 1*: Continuous values drawn from a uniform distribution between 0 and 1
  - *Column 2*: Binary values (0 or 1) drawn from an uneven coin flip where heads (1) has a probability of 0.6, and tails (0) has a probability of 0.4

➤ *Hints: see `np.random.rand()` and `np.random.choice()` examples in Notebook*

# Using Generate AI to Generate Code

## *Prompt*

- Same text as the problem instructions. **Clear and step-by-step instructions** are often enough to get Copilot to generate good code

## *Model*

- The model used to generate the code here is **Claude 3.5 Sonnet**

Generate a 2D NumPy array with 100 rows and 3 columns:

Column 0: row numbers 0-99

Column 1: Continuous values drawn from a uniform distribution between 0 and 1

Column 2: Binary values (0 or 1) drawn from an uneven coin flip where heads (1) has a probability of 0.6, and tails (0) has a probability of 0.4

week3\_notebook\_numpy.ipynb • Cell 1 Current file ⌂

@ ⌂ ⌂

Claude 3.5 Sonnet (Preview) ⌂

▷ ⌂

# Using Generate AI to Generate Code

## *Copilot Output*

- Starts with a description or plan for generating the code
- Then provides the Python code
- *Note: you may get a different response or code, but they should work*

 GitHub Copilot

**Plan:**

1. Import required libraries (numpy)
2. Create array of row numbers (0-99)
3. Generate uniform random numbers between 0-1
4. Generate binary values with 0.6 probability for 1
5. Combine into 2D array
6. Save and display result

**Code:**

```
import numpy as np

# Set random seed for reproducibility
np.random.seed(42)

# Create the array
data = np.zeros((100, 3))

# Column 0: Row numbers
data[:, 0] = np.arange(100)

# Column 1: Uniform distribution between 0-1
data[:, 1] = np.random.uniform(0, 1, 100)

# Column 2: Binary values with p(1)=0.6
data[:, 2] = np.random.choice([0, 1], size=100, p=[0.4, 0.6])

# Print first few rows
print("First 5 rows of the array:")
print(data[:5])

# Save to file
np.save('random_array.npy', data)
```

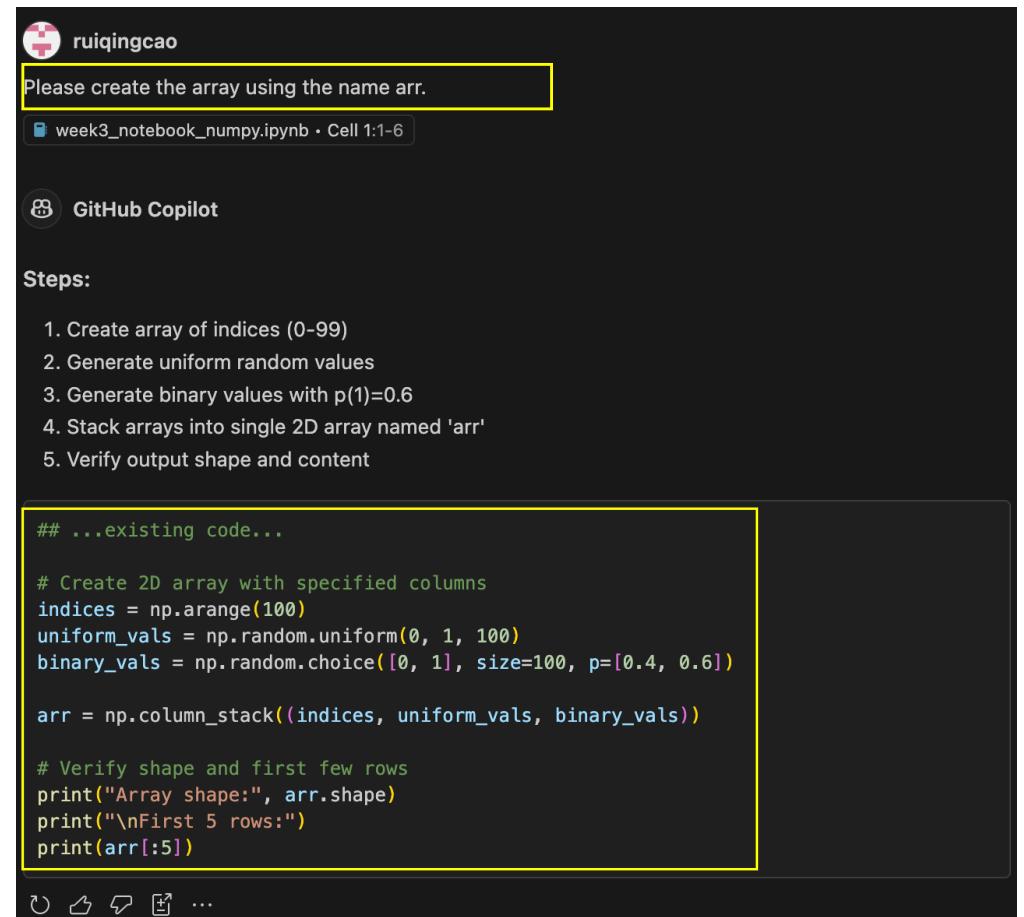
To run:

1. Open terminal (Cmd + `)
2. Run: `python numpy_array.py`

# Using Generate AI to Generate Code

## *Ask for code adjustments*

- “Please create the array using the name arr.”
- Generates Python code again to **meet the new requirement**



ruiqingcao

Please create the array using the name arr.

week3\_notebook\_numpy.ipynb • Cell 1:1-6

GitHub Copilot

Steps:

1. Create array of indices (0-99)
2. Generate uniform random values
3. Generate binary values with  $p(1)=0.6$
4. Stack arrays into single 2D array named 'arr'
5. Verify output shape and content

```
## ...existing code...
# Create 2D array with specified columns
indices = np.arange(100)
uniform_vals = np.random.uniform(0, 1, 100)
binary_vals = np.random.choice([0, 1], size=100, p=[0.4, 0.6])

arr = np.column_stack((indices, uniform_vals, binary_vals))

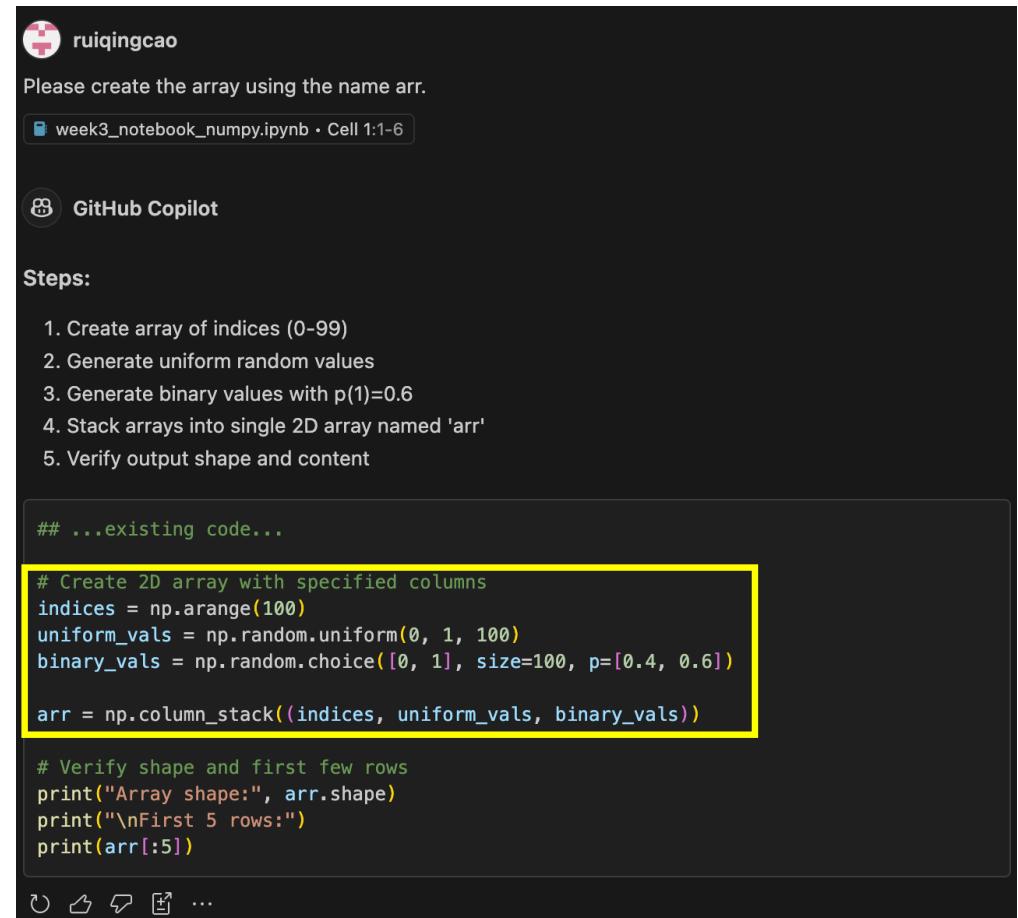
# Verify shape and first few rows
print("Array shape:", arr.shape)
print("\nFirst 5 rows:")
print(arr[:5])
```

undo redo cell magic ...

# Using Generate AI to Generate Code

## *Select generated code partially*

- Sometimes you only need part (not all) of the generated code
- Understand the meaning of the code, and select the part of the code that does what you need
- For example, no need to print the first 5 rows of the array



ruiqingcao  
Please create the array using the name arr.  
week3\_notebook\_numpy.ipynb • Cell 1:1-6

GitHub Copilot

Steps:

1. Create array of indices (0-99)
2. Generate uniform random values
3. Generate binary values with  $p(1)=0.6$
4. Stack arrays into single 2D array named 'arr'
5. Verify output shape and content

```
## ...existing code...
# Create 2D array with specified columns
indices = np.arange(100)
uniform_vals = np.random.uniform(0, 1, 100)
binary_vals = np.random.choice([0, 1], size=100, p=[0.4, 0.6])

arr = np.column_stack((indices, uniform_vals, binary_vals))

# Verify shape and first few rows
print("Array shape:", arr.shape)
print("\nFirst 5 rows:")
print(arr[:5])
```

# Indexing a Numpy Array

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- Recall that `list` is a Python built-in type: a NumPy array is very similar to a `list`, but has more capabilities and computes faster
- Recall that indexes **start at 0**: this applies to *all array-like data objects*, including NumPy arrays
  - Index of a 1D array `a`: `a[n]` selects the element located at `n` (in `a`)
  - Index of a 2D array `b`: `b[m][n]` or `b[m, n]` selects the element located at row `m` and column `n` (in `b`)

# Slicing a NumPy Array

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General syntax: `array[start:end:step]` (similar to list)

Slice the 1D array `a`:

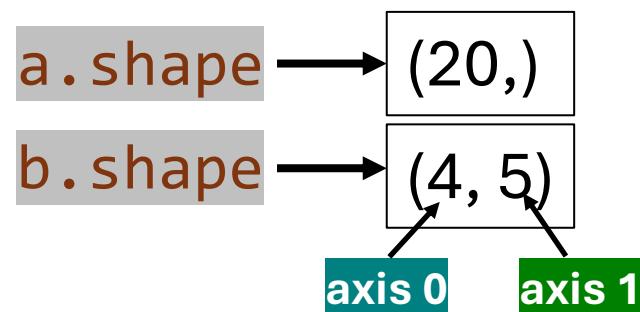
- Select elements from *index n to m*: `a[n:m+1]`
- Select elements from *index n to the end*: `a[n:]`
- Select the *first n elements* (from 0 to n-1): `a[:n]`

Slice the 2D array `b`:

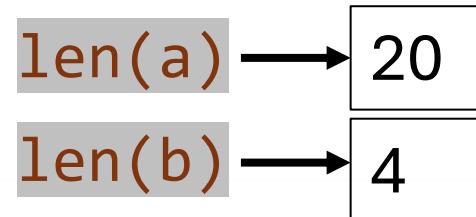
- Select *row m* as a 1D array: `b[m, :]`
- Select *column n* as a 1D array: `b[:, n]`

# Size of a NumPy Array

- The `ndarray.shape` attribute returns the dimension of an array as a tuple. Suppose `a` is a 1D array of length 20, and `b` is a 2D array with 4 rows and 5 columns



- Alternatively, you can use the built-in `len()` function to return the number of elements along the first dimension (axis 0) of an object



# Sorting a Numpy Array

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The `np.sort()` function can be used to sort arrays

- Sort a 1D array `a`:

```
np.sort(a)
```

- Sort a 2D array `b` on its values in column j (note: `np.lexsort()` takes a list of columns to sort on from least to most important)

```
np.lexsort([b[:,j]])
```

Note: `np.sort()` always sorts numbers in ascending order; To reverse the order, simply apply slicing after sorting: e.g., `np.sorted(a)[::-1]`

# Unique Values & Frequencies Counts

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- A very common way to explore a particular variable in a data set is to look at its range (i.e., unique values that it can possibly take)

```
np.unique(a)
```

...returns the unique values in the 1D array **a**

- Relatedly, we may also want to know the frequency of each value

```
np.unique(a, return_counts=True)
```

...returns unique values and their frequency counts in the 1D array **a**

# Summary Statistics of Data in 2D Array

A 2D NumPy array **b** stores tabular data, with rows as observations and columns as variables. To compute summary statistics for each variable, aggregate the array along rows (axis=0) using aggregation functions like mean, standard deviation, median, max, or min

Statistic	Aggregation Function	Robust to Missing Values
Mean	<code>np.mean(b, axis=0)</code>	
StDev	<code>np.std(b, axis=0)</code>	
Median	<code>np.median(b, axis=0)</code>	
Maximum	<code>np.max(b, axis=0)</code>	
Minimum	<code>np.min(b, axis=0)</code>	
Sum	<code>np.sum(b, axis=0)</code>	

Problem: returns `np.nan` if there is even one missing value in a variable

# Summary Statistics of Data in 2D Array

A 2D NumPy array **b** stores tabular data, with rows as observations and columns as variables. To compute summary statistics for each variable, aggregate the array along rows (axis=0) using aggregation functions like mean, standard deviation, median, max, or min

Statistic	Aggregation Function	Robust to Missing Values
Mean	<code>np.mean(b, axis=0)</code>	<code>np.nanmean(b, axis=0)</code>
StDev	<code>np.std(b, axis=0)</code>	<code>np.nanstd(b, axis=0)</code>
Median	<code>np.median(b, axis=0)</code>	<code>np.nanmedian(b, axis=0)</code>
Maximum	<code>np.max(b, axis=0)</code>	<code>np.nanmax(b, axis=0)</code>
Minimum	<code>np.min(b, axis=0)</code>	<code>np.nanmin(b, axis=0)</code>
Sum	<code>np.sum(b, axis=0)</code>	<code>np.nansum(b, axis=0)</code>

Problem: returns `np.nan` if there is even one missing value in a variable

# Plot Frequency Histogram for a Variable

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- Import the visualization library first

```
from matplotlib import pyplot as plt
```

- The underlying NumPy function `np.histogram()` generates frequency histograms, but for plotting, it does not need to be directly called. Instead

```
plt.hist(array[:,j])  
plt.show()
```

...plots the frequency histogram of the variable in array's column j

# Exercise: NumPy Array Basic Operations

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1. Create two 2D NumPy arrays  $X_1$  and  $X_2$  independently using the previous exercise with 100 rows and 3 columns (column 0: 0-99; column 1:  $Uniform[0,1]$ , column 2:  $Bernoulli(0.6)$ )
2. Compute the difference between  $X_1$  and  $X_2$ , square each element of the result, and store the new array in  $Y$   
➤ Hint: matrix operations work on NumPy arrays (though they don't work on lists)
3. Sort rows of  $Y$  on the values of column 1 (hint: use `np.lexsort()`)
4. Compute the unique value counts for  $Y$ 's column 2
5. Slice rows 50–99 of columns 1 and 2 from  $Y$  into a new array  $Y_2$
6. For each column of  $Y_2$ , compute the mean, median, and standard deviation, and plot the frequency histograms