



Understanding Python, NumPy, and Tensors

Further Considerations

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import numpy as np

So, what are tensors?

- A building block of linear algebra
- A mathematical formalism for a multi-dimensional collection / object – houses items, notably numbers / values
 - In Python, there are often called arrays or n-dimensional arrays (*ndarrays*)

import numpy as np

So, what are tensors?

- A building block of linear algebra
- Let's revisit numbers
 and objects collections
 of them. onal collection / A mathematical formalism for a permanent object - houses items, note
 - In Python, there nal arrays (*ndarrays*)

Scalars

- A scalar is a single number (think of it as an "atomic" object)
- Examples: Integers, real numbers, rational numbers, etc.
- Denoted with italic font:

On number spaces / domains:

 $\mathbb{R}\mathcal{R}$ All real (continuous) numbers

 $\mathbb{Z}\mathcal{Z}$ All integers

All natural (counting) numbers

Vectors

- An array of numbers arranged in order
- Each no. identified by an index
- Written in lower-case bold such as x
 - its elements are in italics lower case, subscripted

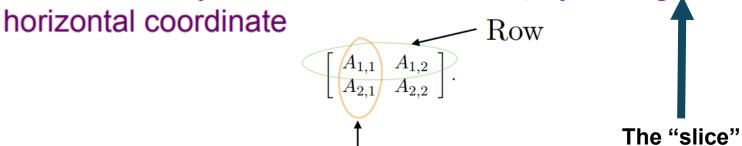
$$oldsymbol{x} = egin{bmatrix} x_1 \ x_2 \ dots \ x_n \end{bmatrix}$$
 $oldsymbol{\mathcal{R}}^n = oldsymbol{\mathcal{R}}^{n imes 1}$ "All real numbers"

- If each element is in R then \boldsymbol{x} is in R^n
- We can think of vectors as points in space
 - Each element gives coordinate along an axis

Matrices

- 2-D array of numbers
 - So each element identified by two indices
- Denoted by bold typeface A
 - Elements indicated by name in italic but not bold
 - $A_{1,1}$ is the top left entry and $A_{m,n}$ is the bottom right entry

• We can identify nos in vertical column j by writing : for the



Column

- $A_{i:}$ is i^{th} row of A, $A_{:j}$ is j^{th} column of \boldsymbol{A}
- If A has shape of height m and width n with real-values then $A \in \mathbb{R}^{m \times n}$

operator

Tensors

- Sometimes need an array with more than two axes
 - E.g., an RGB color image has three axes
- A tensor is an array of numbers arranged on a regular grid with variable number of axes
- Denote a tensor with this bold typeface: A

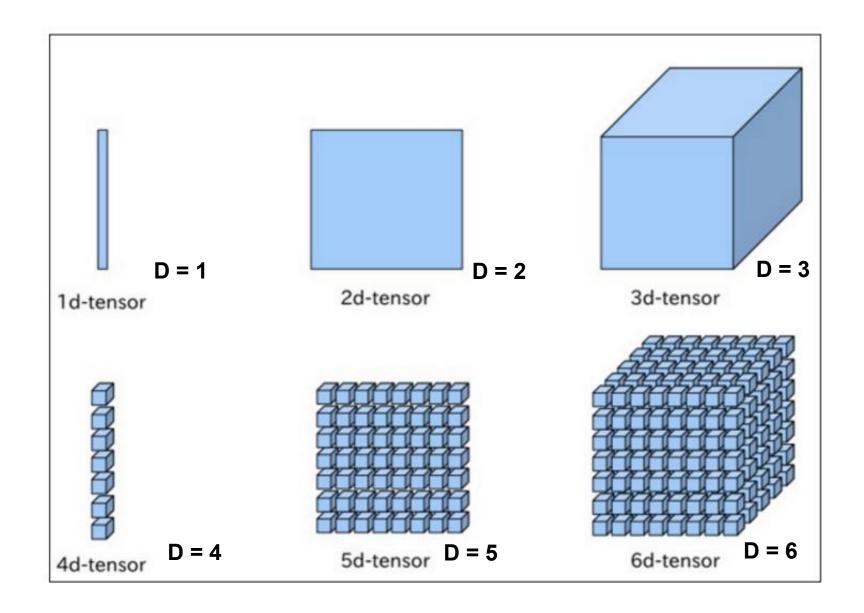
3D tensor

• Element (i,j,k) of tensor denoted by $A_{i,j,k}$

Types of notation accepted (just be consistent & mean what you write):

$$\mathcal{R}^{1 imes1 imes1}=\mathcal{R}^{1 imes1}=\mathcal{R}^1=\mathcal{R}$$

Shapes of Tensors



The ndarray (the Python tensor)

Table 2-1. Basic Attributes of the ndarray Class

Attribute	Description	
shape	A tuple that contains the number of elements (i.e., the length) for each dimension (axis) of the array	
size	The total number of elements in the array	
ndim	The number of dimensions (axes)	
nbytes	The number of bytes used to store the data	
dtype	The data type of the elements in the array	

- The ndarray (object) contains a lot of useful metadata that you can access
 - Example: ndarray.shape > returns a tuple/pair with dimension values
- Array ordering row ("C") versus column ("F" for Fortran) major

When in doubt, read the manual for an object:

```
>>> data = np.array([[1., 2.], [3., 4.]], dtype=np.float32)
>>> help(data)
>>>
```

Data Types

Table 2-2. Basic Numerical Data Types Available in NumPy

dtype	Variants	Description
int	int8, int16, int32, int64	Integers
uint	uint8, uint16, uint32, uint64	Unsigned (nonnegative) integers
bool	bool	Boolean (True or False)
float	float16, float32, float64, float128	Floating-point numbers
complex	complex64, complex128, complex256	Complex-valued floating-point numbers

- You will primarily work with floating-point (like float32), integer (int32), and Boolean (bool) in this course
- Become comfortable with type-casting
- Become comfortable with type promotion

```
In [3]: type(data)
Out[3]: numpy.ndarray
In [4]: data
Out[4]: array([[1, 2],
              [3, 4],
               [5, 6]])
In [5]: data.ndim
Out[5]: 2
In [6]: data.shape
Out[6]: (3, 2)
In [7]: data.size
Out[7]: 6
In [8]: data.dtype
Out[8]: dtype('int64')
In [9]: data.nbytes
Out[9]: 48
```

Creating Different Kinds of (Initial) Arrays

 Useful ways to create (instantiate) ndarrays pre-filled with particular values

np.eye: Creates diagonal array with values on main diagonal

Function Name	Type of Array
np.array	Create an array for which the elements are given by an array-like object, which, for example, can be a (nested) Python list, a tuple, an iterable sequence, or another ndarray instance.
np.zeros	Create an array with the specified dimensions and data type that is filled with zeros.
np.ones	Create an array with the specified dimensions and data type that is filled with ones.
np.diag	Create a diagonal array with specified values along the diagonal and zeros elsewhere.
np.arange	Create an array with evenly spaced values between the specified start, end, and increment values.
np.linspace	Create an array with evenly spaced values between specified start and end values, using a specified number of elements.
np.logspace	Create an array with values that are logarithmically spaced between the given start and end values.
np.meshgrid	Generate coordinate matrices (and higher-dimensional coordinate arrays) from one-dimensional coordinate vectors.
np.fromfunction	Create an array and fills it with values specified by a given function, which is evaluated for each combination of indices for the given array size.
np.fromfile	Create an array with the data from a binary (or text) file. NumPy also provides a corresponding function np.tofile with which NumPy arrays can be stored to disk and later read back using np.fromfile.
np.genfromtxt, np.loadtxt	Create an array from data read from a text file, for example, a comma-separated value (CSV) file. The np.genfromtxt function also supports data files with missing values.
np.random.rand	Generate an array with random numbers that are uniformly distributed between 0 and 1. Other types of distributions are also available in the np.random module.

Questions?

