

Deep Learning: Introduction to Numpy

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- 1 Introduction to numpy

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- 2 Simple computational graph

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- ➊ Introduction to numpy
- ➋ Simple computational graph
- ➌ Example of gradient descent

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- ➊ Introduction to numpy
- ➋ Simple computational graph
- ➌ Example of gradient descent
- ➍ Logistic Regression from scratch
- ➎ Introduction to deep learning frameworks (PyTorch)

Section 1

Introduction to numpy

What is NumPy?

- NumPy is a Python library used for working with arrays

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- Very often used for linear algebra, fourier transformations and matrices

What is NumPy?

- NumPy is a Python library used for working with arrays
- Very often used for linear algebra, fourier transformations and matrices
- NumPy stands for Numerical Python

Benefits of using NumPy

- In Python we have lists that serve the purpose of arrays, but they are slow to process.

Benefits of using NumPy

- In Python we have lists that serve the purpose of arrays, but they are slow to process.
- NumPy aims to provide an array object that is up to 50x faster than traditional Python lists.

Importing numpy

NumPy is usually imported under the np alias.

```
import numpy as np  
  
print(np.__version__)  
  
## 1.21.5
```

Arrays in numpy

```
x = np.array([10, 1, 1, 2, 3])  
print(x)
```

```
## [10  1  1  2  3]
```

```
print(x.shape)
```

```
## (5,)
```

The shape of array

- For 1D array, return a shape tuple with only 1 element (i.e. $(n,)$)

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- For 1D array, return a shape tuple with only 1 element (i.e. $(n,)$)
- For 2D array, return a shape tuple with only 2 elements (i.e. (n,m))
- For 3D array, return a shape tuple with only 3 elements (i.e. (n,m,k))

The shape of array

- For 1D array, return a shape tuple with only 1 element (i.e. $(n,)$)
- For 2D array, return a shape tuple with only 2 elements (i.e. (n,m))
- For 3D array, return a shape tuple with only 3 elements (i.e. (n,m,k))
- For 4D array, return a shape tuple with only 4 elements (i.e. (n,m,k,j))

Introduction to numpy

Creating a column vector

```
x = np.ones(5)
print(x)

## [1.  1.  1.  1.  1.]

print(x.shape)

## (5,)

print(x.shape)

## (5,)

x = x.reshape(5, 1)
print(x)

## [[1.]
##  [1.]
##  [1.]
##  [1.]
##  [1.]]
```

Creating a column vector

Same as

```
x = np.ones(5).reshape(-1,1)
print(x.shape)

## (5, 1)
```

Creating a row vector

```
x = x.reshape(1,5)  
print(x)
```

```
## [[1. 1. 1. 1. 1.]]
```

n-d arrays

Create 2d array

```
x = np.array([[1,5,2,3], [5,6,3,3]])  
print(x)
```

```
## [[1 5 2 3]  
##  [5 6 3 3]]
```

n-d arrays

Is this 2d or 3d array?

```
x = np.array([[1,5,3,2],[2,2,2,2],[3,5,3,3]])
```

n-d arrays

```
x = np.array([[[1,5,3,2], [2,2,2,2], [3,5,3,3]],  
             [[1,5,7,2], [1,2,6,2], [1,2,6,2]]])  
print(x)
```

```
## [[1 5 3 2]  
##   [2 2 2 2]  
##   [3 5 3 3]]  
##  
## [[1 5 7 2]  
##   [1 2 6 2]  
##   [1 2 6 2]]
```


Indexing

```
x = np.array([10,1,1,2,3])  
# the fourth element from the array  
print(x[3])  
  
## 2  
  
# Everything up to the second element  
print(x[:2])  
  
## [10  1]
```

Introduction to numpy

Indexing n-d arrays

```
x = np.arange(21)
print(x)
```

```
## [ 0  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20]
```

```
x = x.reshape(3, 7)
print(x)
```

```
## [[ 0  1  2  3  4  5  6]
##   [ 7  8  9 10 11 12 13]
##   [14 15 16 17 18 19 20]]
```

```
# Second element of the first array
print(x[0,1])
```

```
## 1
```

```
# First two rows of the array
print(x[:2,:])
```

```
## [[ 0  1  2  3  4  5  6]
##   [ 7  8  9 10 11 12 13]]
```

Indexing n-d arrays

```
# First two columns of the array
```

```
print(x[:,2])
```

```
## [[ 0  1]
```

```
##  [ 7  8]
```

```
## [14 15]]
```

```
# first two rows and first two columns
```

```
print(x[:2,:2])
```

```
## [[0 1]
```

```
##  [7 8]]
```

```
# Slicing with the range
```

```
print(x[:2,2:5])
```

```
## [[ 2  3  4]
```

```
##  [ 9 10 11]]
```

Numpy methods

```
x = np.array([10, 1, 1, 2, 3])  
print(x.mean())
```

```
## 3.4
```

```
print(x.sum())
```

```
## 17
```

```
print(x.min())
```

```
## 1
```

```
print(x.max())
```

```
## 10
```

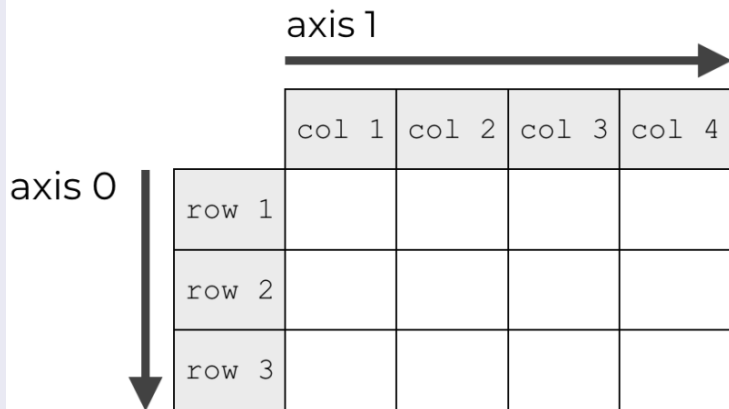
Numpy methods

```
l1 = [5, 2, 3, 4]  
np.mean(l1)
```

```
## 3.5
```

```
# l1.mean() this will rais an error
```


Axes in numpy



Axes in numpy

When you set `axis = 1`, you are doing calculations along the axis 1

axis 1

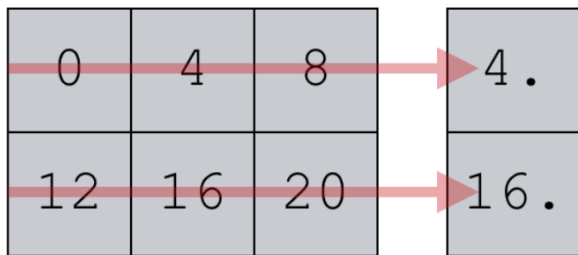


axis 0

0	4	8
12	16	20

Axes in numpy

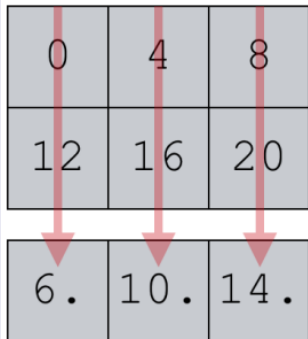
`np.mean` calculates in the column direction when we set `axis = 1`



Axes in numpy

```
print(x.mean(axis = 0))
```

```
## 3.4
```



0	4	8
12	16	20
6.	10.	14.

`np.mean` calculates the mean across the rows when we set `axis = 0`

NumPy basic operations

Generate two random vectors

```
np.random.seed(10)
x = np.random.random(size = 10)
print(x)
```

```
## [0.77132064 0.02075195 0.63364823 0.74880388 0.49850701 0.22479665
##  0.19806286 0.76053071 0.16911084 0.08833981]
```

```
np.random.seed(10)
y = np.random.random(size = 10)
print(y)
```

```
## [0.77132064 0.02075195 0.63364823 0.74880388 0.49850701 0.22479665
##  0.19806286 0.76053071 0.16911084 0.08833981]
```

NumPy basic operations: Element wise operations

```
print(x + y)
```

```
## [1.54264129 0.0415039  1.26729647 1.49760777 0.99701402 0.44959329  
##  0.39612573 1.52106142 0.33822167 0.17667963]
```

```
print(2*x)
```

```
## [1.54264129 0.0415039  1.26729647 1.49760777 0.99701402 0.44959329  
##  0.39612573 1.52106142 0.33822167 0.17667963]
```

```
print(x*y)
```

```
## [5.94935535e-01 4.30643402e-04 4.01510086e-01 5.60707255e-01  
##  2.48509241e-01 5.05335318e-02 3.92288984e-02 5.78406964e-01  
##  2.85984750e-02 7.80392277e-03]
```

NumPy basic operations: Dot product

$$\begin{pmatrix} a_x \\ a_y \\ a_z \end{pmatrix} \bullet \begin{pmatrix} b_x \\ b_y \\ b_z \end{pmatrix} = a_x \cdot b_x + a_y \cdot b_y + a_z \cdot b_z$$

```
print(y.dot(x))
```

```
## 2.510664551824326
```

```
print(np.dot(x, y))
```

```
## 2.510664551824326
```

NumPy basic operations: Matrix multiplication

```
m1 = np.ones(16).reshape(4,4)
np.random.seed(10)
m2 = np.random.rand(16).reshape(4,4)
```

This is not matrix multiplication

```
print(m1*m2)
```

```
## [[0.77132064 0.02075195 0.63364823 0.74880388]
##   [0.49850701 0.22479665 0.19806286 0.76053071]
##   [0.16911084 0.08833981 0.68535982 0.95339335]
##   [0.00394827 0.51219226 0.81262096 0.61252607]]
```

This is the matrix multiplication

```
print(m1.dot(m2))
```

```
## [[1.44288676 0.84608067 2.32969188 3.07525401]
##   [1.44288676 0.84608067 2.32969188 3.07525401]
##   [1.44288676 0.84608067 2.32969188 3.07525401]
##   [1.44288676 0.84608067 2.32969188 3.07525401]]
```

Concatenating numpy arrays

- `np.concatenate()`

```
np.random.seed(10)
ar1 = np.random.randn(20)
ones = np.ones(20)
```

```
ar2 = np.array([ar1, ones])
print(ar2)
```

```
## [[ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556
##      0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737
##     -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688
##      1.484537  -1.07980489]
##  [ 1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         ]]
```

Concatenating numpy arrays

- `np.concatenate()`
- `np.stack()`

```
np.random.seed(10)
ar1 = np.random.randn(20)
ones = np.ones(20)
```

```
ar2 = np.array([ar1, ones])
print(ar2)
```

```
## [[ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556
##      0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737
##     -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688
##      1.484537  -1.07980489]
##  [ 1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         ]]
```

Concatenating numpy arrays

- `np.concatenate()`
- `np.stack()`
- `np.vstack()`

```
np.random.seed(10)
ar1 = np.random.randn(20)
ones = np.ones(20)
```

```
ar2 = np.array([ar1, ones])
print(ar2)
```

```
## [[ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556
##      0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737
##     -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688
##      1.484537  -1.07980489]
##  [ 1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         ]]
```


Concatenating numpy arrays

- `np.concatenate()`
- `np.stack()`
- `np.vstack()`
- `np.hstack()`

```
np.random.seed(10)
ar1 = np.random.randn(20)
ones = np.ones(20)
```

```
ar2 = np.array([ar1, ones])
print(ar2)
```

```
## [[ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556
##      0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737
##     -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688
##      1.484537  -1.07980489]
##  [ 1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         ]]
```

Concatenating numpy arrays

- `np.concatenate()`
- `np.stack()`
- `np.vstack()`
- `np.hstack()`
- etc. . .

```
np.random.seed(10)
ar1 = np.random.randn(20)
ones = np.ones(20)
```

```
ar2 = np.array([ar1, ones])
print(ar2)
```

```
## [[ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556
##      0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737
##     -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688
##      1.484537  -1.07980489]
##  [ 1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         1.         1.         1.         1.
##      1.         1.         ]]
```

Vertical stacking

```
print(np.vstack([ar1, ones]))
```

```
## [[ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556
##      0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737
##     -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688
##      1.484537  -1.07980489]
##  [ 1.         1.         1.         1.         1.         1.
##     1.         1.         1.         1.         1.         1.
##     1.         1.         1.         1.         1.         1.
##     1.         1.         ]]
```

```
ar3 = np.vstack([ar1, ones, np.zeros(20)])
print(ar3)
```

```
## [[ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556
##      0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737
##     -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688
##      1.484537  -1.07980489]
##  [ 1.         1.         1.         1.         1.         1.
##     1.         1.         1.         1.         1.         1.
##     1.         1.         1.         1.         1.         1.
##     1.         1.         ]
##  [ 0.         0.         0.         0.         0.         0.
```

Horizontal stacking

```
ar4 = np.hstack([ar1, ones])  
print(ar4)
```

```
## [ 1.3315865  0.71527897 -1.54540029 -0.00838385  0.62133597 -0.72008556  
##  0.26551159  0.10854853  0.00429143 -0.17460021  0.43302619  1.20303737  
## -0.96506567  1.02827408  0.22863013  0.44513761 -1.13660221  0.13513688  
##  1.484537 -1.07980489  1.          1.          1.          1.  
##  1.          1.          1.          1.          1.          1.  
##  1.          1.          1.          1.          1.          1.  
##  1.          1.          1.          1.          ]
```

Introduction to numpy

np.concatenate()

This one is similar to np.vstack() and np.hstack()

```
ar1 = ar1.reshape(-1,1)
print(ar1.shape)
```

```
## (20, 1)
```

```
ones = ones.reshape(-1,1)
print(ones.shape)
```

```
## (20, 1)
```

```
print(np.concatenate([ar1, ones], axis = 0).shape)
```

```
## (40, 1)
```

```
print(np.concatenate([ar1, ones], axis = 1).shape)
```

```
## (20, 2)
```

```
print(np.vstack([ar1, ones]).shape)
```

```
## (40, 1)
```