

# Preliminaries

Deep learning - Spring 2020

# Preliminaries

## (Python, NumPy)

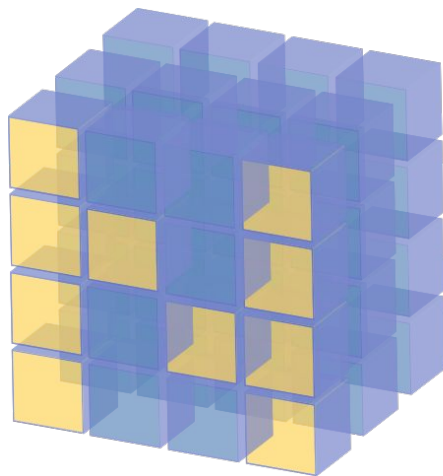
# Why Python?

- Easy-to-use language
- Great community participation
- Decent library availability (especially machine learning libraries)



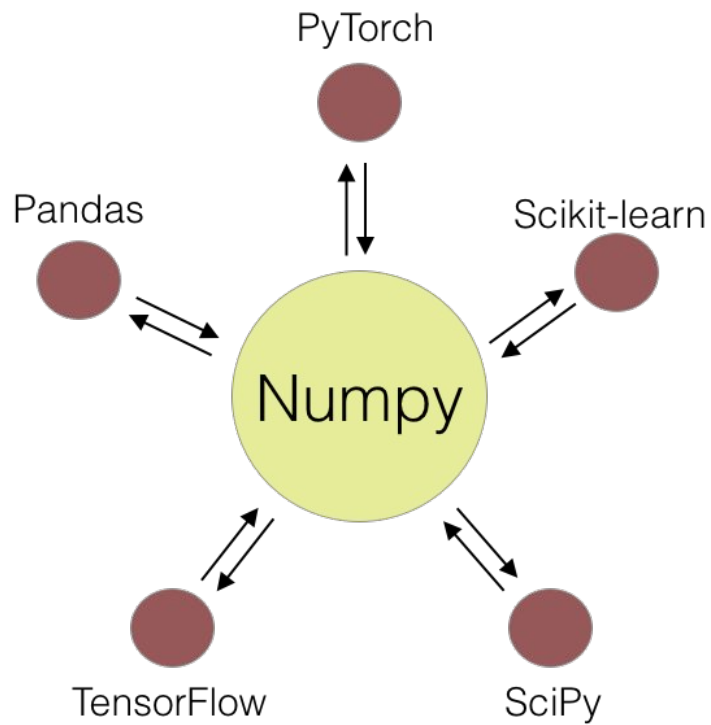
# Numpy

- Core library for scientific computing in Python
- Appropriate for processing homogeneous multidimensional arrays and matrices



NumPy

# Numpy



# Numpy

## Array

- A numpy array is a grid of values, all of the same type
- number of dimensions is the *ndim (rank)* of the array
- The *shape* of an array is a tuple of integers giving the size of array along each dimension
- We can initialize numpy arrays from nested Python lists, and access elements using square brackets

# Numpy

- Numpy is Fast
- Numpy arrays are densely packed arrays of homogeneous type ([locality of reference](#))
- Many Numpy operations are implemented in C
- e.g. if you are summing up two arrays the addition will be performed with the specialized CPU vector operations

➤ [See the example of this part in the notebook]

# Numpy

- Uses much less memory to store data (compared to Python lists)

```
np.array([1, 2, 3, 4], dtype=np.int8)
```

- Simple and beautiful API
- **Indexing**
- **Broadcasting**: Working with arrays of different shapes

➤ [See the examples for this part in the notebook]



# Numpy

- Ex. Linear regression in Numpy

model



$$h(x) = \sum_{i=0}^d \theta_i x_i = \theta^T x$$

loss



$$J(\theta) = \frac{1}{2} \sum_{i=1}^n (h_{\theta}(x^{(i)}) - y^{(i)})^2$$

- LMS algorithm

Repeat until convergence {

$$\theta_j := \theta_j + \alpha \sum_{i=1}^n (y^{(i)} - h_{\theta}(x^{(i)})) x_j^{(i)} \quad (\text{for every } j)$$

}

- The Normal Equation

$$X^T X \theta = X^T \vec{y} \quad \longrightarrow \quad \theta = (X^T X)^{-1} X^T \vec{y}$$

# Links and references

- You can find a great tutorial about Python and Numpy [here](#).
- Stanford CS229 lecture note about regression (and classification) is [here](#).