Deep Learning and PyTorch Tutorial

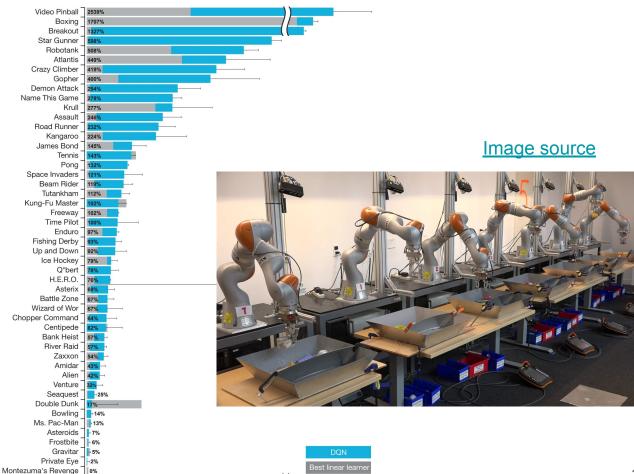
17th Jan 2023 Skanda Vaidyanath CS 234

Overview

- Motivation
- Neural networks
- Backpropagation
- Training and Optimization
- CNNs
- PyTorch

Motivation





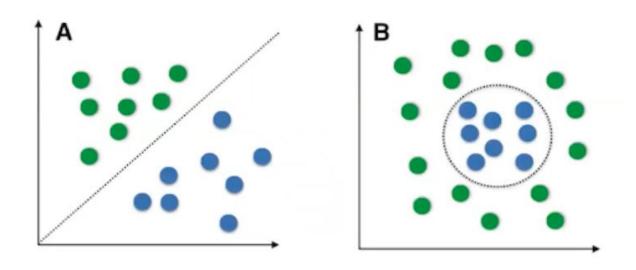
300

400

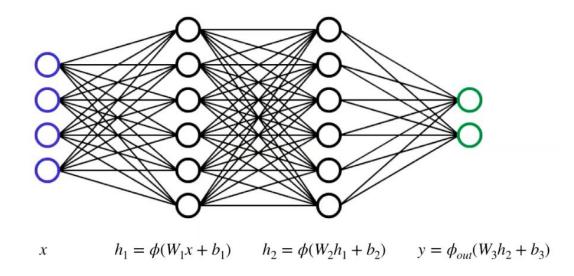
600 1,000

4,500%

The need for non-linear models



Neural networks



Non-linearities

Activation Functions

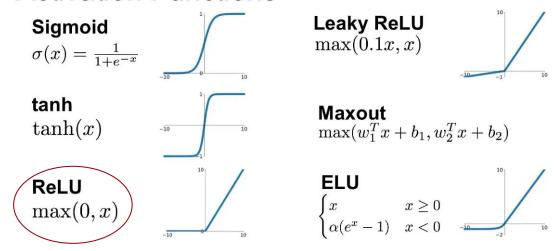


Image credit: Medium article

Computational graphs

Eg: f(x,y,z) = (x+y)z

What is the derivative of f wrt x, y, z?

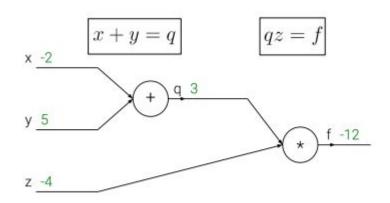
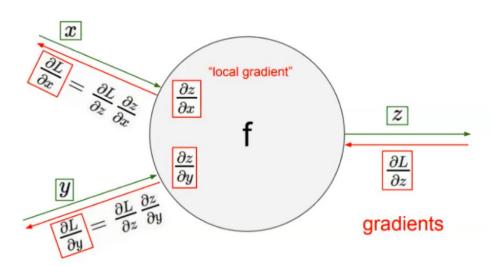


Image credit: <u>CS231N Slides</u>

Backpropagation

Using the chain rule to calculate gradients



Training a neural network

Loss function, usually an average over data points:

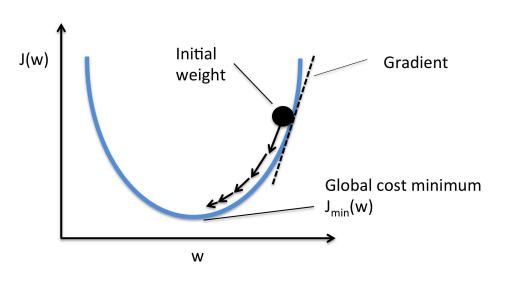
$$L(\theta) = \frac{1}{|\mathcal{D}|} \sum_{(x,y) \in \mathcal{D}} \ell(\theta; x, y)$$

Want to find

$$\theta^* \in \arg\min_{\theta} L(\theta)$$

Usually MSE loss or cross-entropy loss, etc.

Optimization



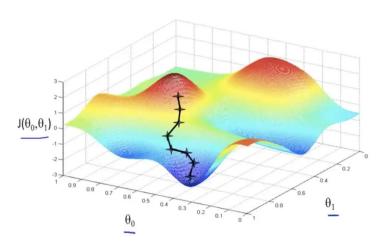


Image credit: Blog post

Image credit: Medium article

Stochastic Gradient Descent

Often it is infeasible/ inefficient to do full batch gradient descent, so we resort to stochastic gradient descent (SGD)

1. Pick some θ_0 , then iterate:

2.
$$\theta_{k+1} = \theta_k - \alpha_k \nabla L(\theta_k)$$

$$\nabla L(\theta) = \frac{1}{|\mathcal{D}|} \sum_{(x,y) \in \mathcal{D}} \nabla_{\theta} \mathcal{E}(\theta; x, y) \qquad O(|\mathcal{D}|)$$

SGD: sample mini-batch $B \subseteq \mathcal{D}$, use

$$\hat{\nabla}L(\theta) = \frac{1}{|B|} \sum_{(x,y) \in B} \nabla_{\theta} \mathcal{E}(\theta; x, y) \qquad O(|B|)$$

Note
$$\mathbb{E}[\hat{\nabla}L(\theta)] = \nabla L(\theta)$$

Other optimizers

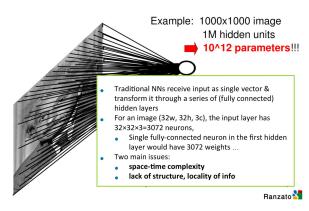
There are several new optimizers that improve on vanilla gradient descent. Some of them are:

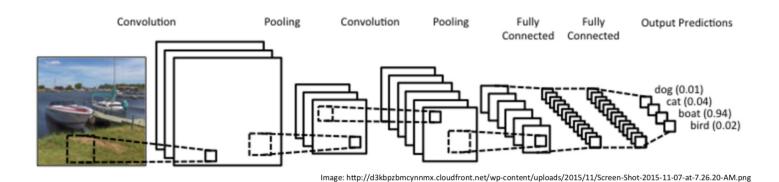
- Momentum
- Adagrad
- RMSProp
- Adam

To learn more: https://ruder.io/optimizing-gradient-descent/

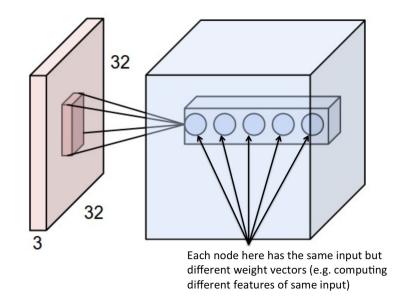
- CNNs are extensively used in computer vision
- If we want to go from pixels to decisions, likely useful to leverage insights from visual input
- Why not just use fully connected networks?





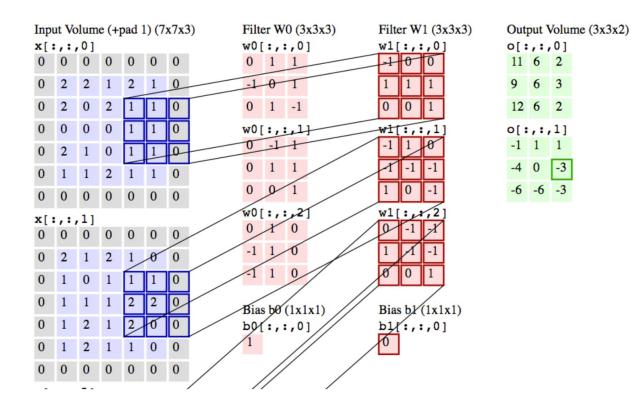


- Image dimensions
- Filters/ kernels
- Stride
- Padding

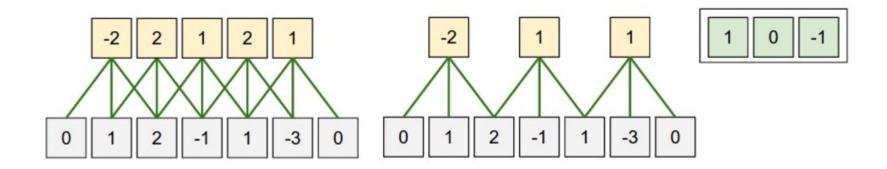


Full demo:

https://cs231n.github .io/convolutional-net works/



Simpler example:



PyTorch

O PyTorch

- Objective: introduce key PyTorch concepts for CS234
- PyTorch documentation: https://pytorch.org/docs/stable/index.html
- Tutorial: Python notebook will be shared after class