

# Training Neural Networks

## • Activation Function

Sigmoid  $\sigma(x) = \frac{1}{1+e^{-x}}$

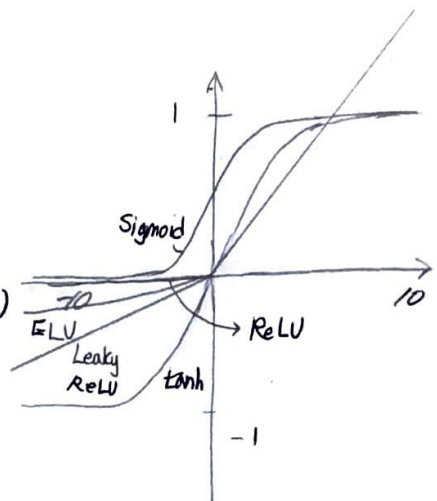
Leaky ReLU  $\max(0.1x, x)$

tanh

Maxout  $\max(w_1^T x + b_1, w_2^T x + b_2)$

ReLU  $\max(0, x)$

ELU  $\begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$



(use ReLU)

## • Data Preprocessing

Normalization  $\rightarrow$  PCA. Whitening of the data

- just subtract the mean image
- subtract per-channel mean

## • Weight Initialization

Small random numbers : bad at deeper network

zero

bad at backpropagation

random / sqrt

is not better

## • Batch Normalization

usually after FC or Convolutional layers

$$\hat{x}^{(k)} = \frac{x^{(k)} - E[x^{(k)}]}{\sqrt{\text{Var}[x^{(k)}]}}$$

recover  $y^{(k)} = \sqrt{\text{Var}[x^{(k)}]} \hat{x}^{(k)} + E[x^{(k)}]$

## • Babysitting the Learning Process

## • Hyperparameter Optimization

Cross-Validation Strategy

random sample hyperparams, in log space when appropriate

• Optimizations: problems with SGD IR2019 update 2921.

SGD  $\mathbf{x}_{t+1} = \mathbf{x}_t - \alpha \nabla f(\mathbf{x}_t)$

↓

SGD + Momentum  $\mathbf{v}_{t+1} = \rho \mathbf{v}_t + \nabla f(\mathbf{x}_t)$

$$\mathbf{x}_{t+1} = \mathbf{x}_t - \alpha \mathbf{v}_{t+1}$$

SGD + Nesterov Momentum  $\mathbf{v}_{t+1} = \rho \mathbf{v}_t - \alpha \nabla f(\mathbf{x}_t + \rho \mathbf{v}_t)$

$$\mathbf{x}_{t+1} = \mathbf{x}_t + \mathbf{v}_{t+1}$$

AdaGrad

$$\mathbf{h}_{t+1} = \mathbf{h}_t + \frac{\partial}{\partial \mathbf{W}} L \odot \frac{\partial}{\partial \mathbf{W}} L$$

$$\mathbf{W}_{t+1} = \mathbf{W}_t - \eta \frac{1}{\sqrt{\mathbf{h}}} \frac{\partial}{\partial \mathbf{W}} L$$

RMSProp

$$\mathbf{h}_{t+1} = \rho \mathbf{h}_t + (1-\rho) \frac{\partial}{\partial \mathbf{W}} L \odot \frac{\partial}{\partial \mathbf{W}} L$$

Adam  
default

1st momentum  $\mathbf{m}$     2nd momentum  $\mathbf{v}$

$$\mathbf{m}_i \leftarrow \beta_1 \mathbf{m}_0 + (1-\beta_1) \mathbf{g}_i$$

$$\hat{\mathbf{m}}_i \leftarrow \frac{\mathbf{m}_i}{1-\beta_1^i} = \frac{\beta_1 \mathbf{m}_0}{1-\beta_1^i} + \frac{(1-\beta_1) \mathbf{g}_i}{1-\beta_1^i} = 0 + \mathbf{g}_i \quad (\because \mathbf{m}_0 = 0) \quad \text{이를 반복}$$

• Model Ensembles

1. train multiple independent models
2. at test time average their result.

• Regularization

- add term to loss

- L2 regularization
- L1 regularization
- Elastic Net (L1+L2)

- Dropout randomly set neurons to zero

- Data Augmentation

random crops and scales  
horizontal flips