

Assignment 6: Introduction to Neural Networks and Backpropagation

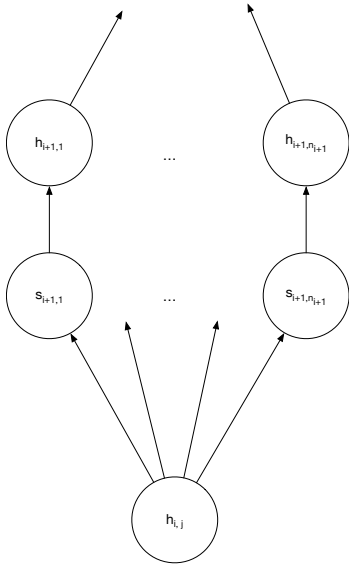
Machine Learning

Fall 2019

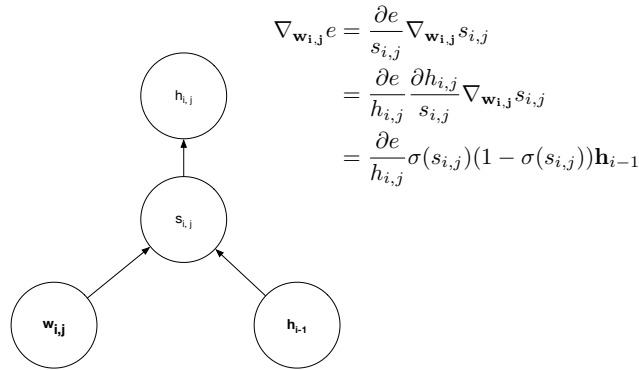
🔗 Learning Objectives

- Gain some familiarity with some of the key ideas in machine learning.
- Review of mathematical concepts we will be using in the beginning part of this course.
- Familiarize yourself with computational tools for machine learning.
- Learn linear regression using a “top-down” approach.

[HMC Multivariable Chain Rule Page](#)

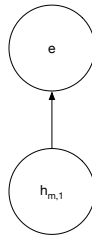


$$\begin{aligned}\frac{\partial e}{\partial h_{i,j}} &= \sum_{k=1}^{n_{i+1}} \frac{\partial s_{i+1,k}}{\partial h_{i,j}} \frac{\partial h_{i+1,k}}{\partial s_{i+1,k}} \\ &= \sum_{k=1}^{n_{i+1}} w_{j,k}^{i+1} \sigma(s_{i+1,k})(1 - \sigma(s_{i+1,k}))\end{aligned}$$



$$\begin{aligned}\nabla_{\mathbf{w}_{i,j}} e &= \frac{\partial e}{\partial s_{i,j}} \nabla_{\mathbf{w}_{i,j}} s_{i,j} \\ &= \frac{\partial e}{\partial h_{i,j}} \frac{\partial h_{i,j}}{\partial s_{i,j}} \nabla_{\mathbf{w}_{i,j}} s_{i,j} \\ &= \frac{\partial e}{\partial h_{i,j}} \sigma(s_{i,j})(1 - \sigma(s_{i,j})) \mathbf{h}_{i-1}\end{aligned}$$

The last hidden layer can be considered the output.
You could also call this z.
This is the math for log loss.



$$\frac{\partial e}{\partial h_{m,1}} = -y \frac{1}{h_{m,1}} - (1 - y) \frac{1}{1 - h_{m,1}}$$

Todo: use a variable for layer instead of just hidden (this would simplify things).

Hidden unit to error:

$$\frac{\partial e}{\partial h_{i,j}} = \sum_{k=1}^{n_{i+1}} \frac{\partial s_{i+1,k}}{\partial h_{i,j}} \frac{\partial h_{i+1,k}}{\partial s_{i+1,k}} \frac{\partial e}{\partial h_{i+1,k}} \quad (1)$$

$$= \sum_{k=1}^{n_{i+1}} w_{j,k}^{i+1} \sigma(s_{i+1,k})(1 - \sigma(s_{i+1,k})) \frac{\partial e}{\partial h_{i+1,k}} \quad (2)$$

Weights to error:

$$\nabla_{\mathbf{w}_{i,j}} e = \frac{\partial e}{\partial s_{i,j}} \nabla_{\mathbf{w}_{i,j}} s_{i,j} \quad (3)$$

$$= \frac{\partial e}{\partial h_{i,j}} \frac{\partial h_{i,j}}{\partial s_{i,j}} \nabla_{\mathbf{w}_{i,j}} s_{i,j} \quad (4)$$

$$= \frac{\partial e}{\partial h_{i,j}} \sigma(s_{i,j})(1 - \sigma(s_{i,j})) \mathbf{h}_{i-1} \quad (5)$$

Output to error (serves as a base case. For simplicity we use $h_{m,1}$ to refer to the single node in the m th layer (which is the output layer).

$$\frac{\partial e}{\partial h_{m,1}} = -y \frac{1}{h_{m,1}} - (1 - y) \frac{1}{1 - h_{m,1}} \quad (6)$$