

→ Back propagation

$$\nabla_{n-1} = \nabla_n * W_{n-1}^T \qquad \nabla = \frac{\partial E}{\partial w}$$

→ Use the chain rule to efficiently compute gradients, top to bottom

$$Error = \frac{1}{2} \sum_i^n (y - \hat{y})^2 \qquad \hat{y} = Sigmoid(x_i \times w_i)$$

$$\frac{\partial E}{\partial w} = \frac{\partial}{\partial w} \frac{1}{2} \sum_i^n (y - \hat{y})^2$$

$$\frac{\partial E}{\partial w} = \sum_i^n (y - \hat{y}) \left(-\frac{\partial E}{\partial w} \hat{y} \right) \qquad \Rightarrow \left(-\frac{\partial E}{\partial w} \hat{y} \right) = \hat{y}(1 - \hat{y})$$