

Mathematische Beschreibung des RNN-Models

Sei X ein Vektor mit n Elementen, wobei n die Anzahl der Unicode Zeichen ist.

Sei Y ein Vektor mit n Elementen, wobei n die Anzahl der Klassifizierungskategorien ist.

So ist Y durch

$$Y = \sigma(W_i(X+H) + B_i)$$

definiert.

Wobei σ durch

$$\sigma_i = \log\left(\frac{e^{x_i}}{\sum_{j=1}^n e^{x_j}}\right)$$

$$\Rightarrow \sigma = \begin{pmatrix} \sigma_1 \\ \sigma_2 \\ \sigma_3 \\ \vdots \end{pmatrix}$$

und H durch

$$H = W_i(X+H) + B_i$$

So kommen wir zu folgendem Optimierungsproblem:

$$J = \left(\sum_{i=1}^N \sigma_{y_{1,i}} \right) / n$$

Die Lösung des Optimierungsproblems erfolgt mit dem Gradientenverfahren:

$$\nabla W_1 = \frac{\partial J}{\partial W_1} = \frac{\partial J}{\partial y} \cdot \frac{\partial y}{\partial W_1}$$

$$\nabla B_1 = \frac{\partial J}{\partial B_1} = \frac{\partial J}{\partial y} \cdot \frac{\partial y}{\partial B_1}$$

$$\nabla W_2 = \frac{\partial J}{\partial W_2} = \frac{\partial J}{\partial y} \cdot \frac{\partial y}{\partial H} \cdot \frac{\partial H}{\partial W_2}$$

$$\nabla B_2 = \frac{\partial J}{\partial B_2} = \frac{\partial J}{\partial y} \cdot \frac{\partial y}{\partial H} \cdot \frac{\partial H}{\partial B_2}$$

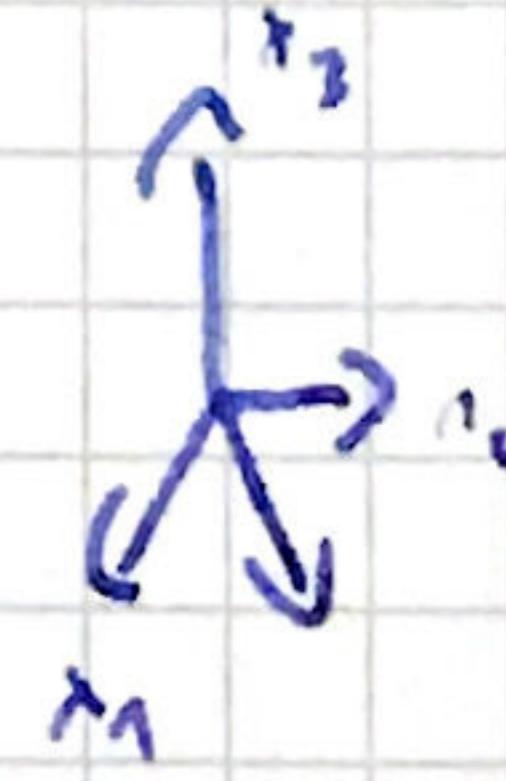
Daraus ergeben sich folgende rekursive Gleichungen

$$W_1 = W_1' - \alpha \cdot \nabla W_1$$

$$W_2 = W_2' - \alpha \cdot \nabla W_2$$

$$B_1 = B_1' - \alpha \cdot \nabla B_1$$

$$B_2 = B_2' - \alpha \cdot \nabla B_2$$



Notizen:

$$\sigma_i = \frac{e^{x_i}}{\sum_{k=1}^K e^{x_k}}$$

$$u(x) = e^{x_i}$$

$$v(x) = \sum_{k=1}^K e^{x_k}$$

$$\sigma_i = \frac{e^{(w_i \cdot (x+h) + b_i)}}{\sum_{k=1}^K e^{(w_i \cdot (x+h) + b_i)}}$$

$$\frac{d\sigma_i}{dx_i} \sigma'_i = \frac{e^{x_i} \cdot \sum_{k=1}^K e^{x_k} - e^{x_i} \cdot e^{x_i}}{\left(\sum_{k=1}^K e^{x_k}\right)^2}$$

$$= \frac{e^{x_i} \cdot \sum_{k=1}^K e^{x_k}}{\left(\sum_{k=1}^K e^{x_k}\right)^2} - \frac{e^{x_i} \cdot e^{x_i}}{\left(\sum_{k=1}^K e^{x_k}\right)^2}$$

$$\frac{d\sigma_i}{dx_i} = \frac{e^{x_i}}{\sum_{k=1}^K e^{x_k}} - \frac{(e^{x_i})^2}{\left(\sum_{k=1}^K e^{x_k}\right)^2}$$

~~$$y_i = \frac{e^{(w_i \cdot (x_i + h_i) + b_i)}}{\sum_{k=1}^K e^{(w_i \cdot (x_k + h_i) + b_i)}}$$~~

$$y(i) = \frac{e^{(w_i \cdot (x_i + h_i) + b_i)}}{\sum_{k=1}^K e^{(w_i \cdot (x_k + h_i) + b_i)}}$$

$$u(x) = e^{(w_i \cdot (x+h) + b_i)}$$

$$\frac{du}{dx_i} u'(x) = e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h)$$

$$v(x) = \sum_{k=1}^K e^{(w_i \cdot (x+h) + b_i)}$$

$$\frac{dv}{dx_i} v'(x) = e^t$$

$$J = \sum_{i=1}^N y(i)$$

$$\frac{dJ}{dy} =$$

$$(u(v(x)))' = u'(v(x)) \cdot v'(x)$$

$$\nabla w_i = \frac{\partial J}{\partial w_i} = \frac{\partial J}{\partial y} \cdot \frac{\partial y}{\partial w_i}$$

~~$$= \frac{\partial J}{\partial y} \cdot \left(\frac{e^{(w_i \cdot (x_i + h_i) + b_i)}}{\sum_{k=1}^K e^{(w_i \cdot (x_k + h_i) + b_i)}} \right)$$~~

$$\frac{du}{dt_i} = e^{(w_i \cdot (x+h) + b_i)} \cdot w_i$$

$$f(x) = w_i \cdot (x+h) + b_i$$

$$= w_i x + w_i h + b_i$$

$$\frac{df}{dh} = w_i$$

$$\frac{dv}{dt_i} = \text{?}$$

$$\frac{\partial Y}{\partial B_i} = \frac{d\sigma_i}{dB_i}$$

$$= \frac{e^{(w_i^j \cdot (x+h) + B_i^j)} \cdot \sum_{k=1}^K e^{(w_i^k \cdot (x+h) + B_i^k)} - e^{(w_i^j \cdot (x+h) \cdot B_i^j)} \cdot e^{(w_i^j \cdot (x+h) \cdot B_i^j)}}{\left(\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + B_i^k)} \right)^2}$$

$$= \frac{e^{(w_i^j \cdot (x+h) + B_i^j)}}{\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + B_i^k)}} - \frac{(e^{(w_i^j \cdot (x+h) \cdot B_i^j)})^2}{\left(\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + B_i^k)} \right)^2}$$

$$\boxed{\frac{\partial Y}{\partial B_i} = \frac{e^{(w_i^j \cdot (x+h) + B_i^j)}}{\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + B_i^k)}} = e^{(w_i^j \cdot (x+h) \cdot B_i^j)} \cdot \lambda}$$

$$\boxed{\frac{\partial Y}{\partial w_i}}$$

$$\frac{\partial Y}{\partial w_i} = \frac{d\sigma}{dw_i} = \frac{e^{(w_i \cdot (x+h) + b_i)}}{\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)}}$$

$$= \frac{e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h) \cdot \sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)} - e^{(w_i \cdot (x+h) + b_i)} \cdot e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h)}{\left(\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)} \right)^2}$$

$$= \frac{e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h)}{\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)}} - \frac{e^{(w_i \cdot (x+h) + b_i)} \cdot e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h)}{\left(\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)} \right)^2}$$

$$= e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h) \cdot \left(\frac{1}{\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)}} - \frac{e^{(w_i \cdot (x+h) + b_i)}}{\left(\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)} \right)^2} \right)$$

$$\frac{\partial Y}{\partial w_i} = e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h) \cdot \left(\frac{1}{\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)}} - \frac{e^{(w_i \cdot (x+h) + b_i)}}{\left(\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)} \right)^2} \right)$$

$$\boxed{\lambda = \frac{1}{\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)}} - \frac{e^{(w_i \cdot (x+h) + b_i)}}{\left(\sum_{k=1}^K e^{(w_i^k \cdot (x+h) + b_i^k)} \right)^2}}$$

$$\boxed{\frac{\partial Y}{\partial w_i} = e^{(w_i \cdot (x+h) + b_i)} \cdot (x+h) \cdot \boxed{\lambda}}$$

$$\frac{\partial Y}{\partial w_i} = \frac{\partial e^{H_i}}{\partial H_i} \cdot \frac{\partial H_i}{\partial w_i}$$

$$\frac{\partial Y}{\partial H_i} = \frac{e^{(w_i \cdot (x+H) + b_i)} \cdot w_i \cdot \sum_{k=1}^K e^{(w_k^k \cdot (x+H) + b_k^k)} - e^{(w_i \cdot (x+H) + b_i)} \cdot e^{(w_j \cdot (x+H) + b_j)}}{\left(\sum_{k=1}^K e^{(w_k^k \cdot (x+H) + b_k^k)}\right)^2}$$

$$\boxed{\frac{\partial Y}{\partial H_i} = e^{(w_i \cdot (x+H) + b_i)} \cdot w_i \cdot \lambda}$$

$$\frac{\partial H_i}{\partial w_i} = (x+H)$$

$$\Rightarrow \boxed{\frac{\partial Y}{\partial w_i} = e^{(w_i \cdot (x+H) + b_i)} \cdot w_i \cdot \lambda \cdot (x+H)}$$

$$\frac{\partial Y}{\partial b_i} = \frac{\partial Y}{\partial H_i} \cdot \frac{\partial H_i}{\partial b_i}$$

$$\frac{\partial H_i}{\partial b_i} \overset{\text{tun}}{=} 1$$

$$\Rightarrow \boxed{\frac{\partial Y}{\partial b_i} = e^{(w_i \cdot (x+H) + b_i)} \cdot w_i \cdot \lambda}$$

$$\nabla W_1^i = \frac{\partial J}{\partial W_i^i} = \boxed{\frac{\partial J}{\partial Y} \cdot e^{(W_i^i \cdot (x+H) + B_i^i)} \cdot (x+H) \cdot \underline{\lambda}}$$

$$\nabla B_1^i = \frac{\partial J}{\partial B_i^i} = \boxed{\frac{\partial J}{\partial Y} \cdot e^{(W_i^i \cdot (x+H) + B_i^i)} \cdot \underline{\lambda}}$$

$$\nabla W_2^i = \frac{\partial J}{\partial W_i^i} = \frac{\partial J}{\partial Y} \cdot \frac{\partial Y}{\partial E} \cdot \frac{\partial E}{\partial H_i} \cdot \frac{\partial H_i}{\partial W_i^i} = \boxed{\frac{\partial J}{\partial Y} \cdot e^{(W_i^i \cdot (x+H) + B_i^i)} \cdot W_i^i \cdot \underline{\lambda} \cdot (x+H)}$$

$$\nabla B_2^i = \frac{\partial J}{\partial B_i^i} = \frac{\partial J}{\partial Y} \cdot \frac{\partial Y}{\partial H_i} \cdot \frac{\partial H_i}{\partial B_i^i} = \boxed{\frac{\partial J}{\partial Y} \cdot e^{(W_i^i \cdot (x+H) + B_i^i)} \cdot W_i^i \cdot \underline{\lambda}}$$

Wobei $\underline{\lambda}$ durch

$$\underline{\lambda} = \frac{1}{\sum_{k=1}^K e^{(W_i^k \cdot (x+H) + B_i^k)}} - \frac{e^{(W_i^i \cdot (x+H) + B_i^i)}}{\left(\sum_{k=1}^K e^{(W_i^k \cdot (x+H) + B_i^k)}\right)^2}$$

definiert ist.