# **Deep Learning Assignment 1**

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## MLP backprop and NumPy implementation

#### 1.1 Analytical derivation of gradients

### Question 1.1 a)

- 1. Cross entropy module  $\frac{\partial L}{\partial x^{(N)}} = -\sum_i t_i \frac{1}{x_i^{(N)}} = \frac{1}{x_{argmax(t)}^{(N)}}$

2. Softmax module 
$$\frac{\partial x_{i}^{(N)}}{\partial \tilde{x}_{j}^{(N)}} = \frac{exp(\tilde{x}_{j}^{(N)})}{\sum_{i=1}^{d_{N}} exp(\tilde{x}_{i}^{(N)})} - (\frac{exp(\tilde{x}_{j}^{(N)})}{\sum_{i=1}^{d_{N}} exp(\tilde{x}_{i}^{(N)})})^{2} \text{ if } i = j$$

$$-\frac{exp(\tilde{x}_{i}^{(N)})}{\sum_{i=1}^{d_{N}} exp(\tilde{x}_{i}^{(N)})} \frac{exp(\tilde{x}_{j}^{(N)})}{\sum_{i=1}^{d_{N}} exp(\tilde{x}_{i}^{(N)})} \text{ if } i \neq j$$

$$= \delta_{ij}(x_{i}^{(N)} - x_{i}^{N2}) + (1 - \delta_{ij})(-x_{i}x_{j})$$

$$\begin{array}{l} \text{3. } \underset{\partial x^{(l < N)}}{\operatorname{ReLU}} \text{module} \\ \frac{\partial x^{(l < N)}}{\partial \tilde{x}^{(l < N)}} = \\ 1 \text{ if } \tilde{x}^{(l)} \geq 0 \\ 0 \text{ if } \tilde{x}^{(l)} < 0 \end{array}$$

$$\begin{array}{l} \text{4. Linear module} \\ \frac{\partial \bar{x}^{(l)}}{\partial x^{(l-1)}} = W^{(l)} \\ \frac{\partial \bar{x}^{(l)}}{\partial W^{(l)}} = x^{(l-1)} \\ \frac{\partial \bar{x}^{(l)}}{\partial b^{(l)}} = 1 \end{array}$$

Question 1.1 b) 

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$$1. \frac{\partial L}{\partial \tilde{x}^{(N)}} = \frac{\partial L}{\partial x^{(N)}} \frac{\partial x^{(N)}}{\partial \tilde{x}^{(N)}} = \sum_{i} t_{i} \frac{1}{x_{i}^{(N)}} \left( \delta_{ij} (x_{i}^{(N)} - x_{i}^{N2}) + (1 - \delta_{ij})(-x_{i}x_{j}) \right)$$

- 32 1.2 NumPy implementation
- 33 2 PyTorch MLP
- 34 **3 Custom Module: Batch Normalization**
- 35 3.1 Automatic differentiation
- 36 4 PyTorch CNN