
CS5787: Exercises 1

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1 Theory: Question 1 [12.5 pts]

- What is the input X ?
There are m samples in the batch with 10 features each so the input shape is $(m, 10)$
- What is the shape of the hidden layer's weight vector W_h , and the shape of its bias vector b_h ?
To transform an $(m, 10)$ input into a hidden layer of 50 neurons we need a shape W_h to have a shape of $(10, 50)$ so that we get a $(m, 50)$ result. After the weight matrix is applied, we then add the bias b_h to each row of W_h , which means that the result needs to be $(1, 50)$
- What is the shape of the output layer's weight vector W_o , and its bias vector b_o ?
 $W_h : (50, 3), b_h : (1, 3)$
- What is the shape of the network's output matrix Y ? $(m, 3)$
- Write an equation that computes the network's output matrix Y as a function of X , W_h , b_h , W_o , and b_o ?
 $Y = W_o a(W_h X + b_h) + b_o$
Where a is the ReLU activation function, $a(x) = \max(0, x)$.

2 Theory: Question 2 [12.5 pts]

For the first layer there are $3 \times 3 \times 3$ parameters in each of the 100 kernels, and then we need bias parameters for each kernel as well. This totals $3 \times 3 \times 3 \times 100 + 100 = 2800$ parameters. In the second layer there are $3 \times 3 \times 100$ parameters in each of the 200 kernels, and then we need bias parameters for each kernel as well. This totals $3 \times 3 \times 100 \times 200 + 200 = 180200$. In the final layer there are $3 \times 3 \times 200$ parameters in each of the 200 kernels, and then we need bias parameters for each kernel as well. This totals $3 \times 3 \times 200 \times 400 + 400 = 720400$. Adding all the layers together we get $2800 + 180200 + 720400 = 903400$ total parameters.

3 Theory: Question 3 [25 pts]

- $$\begin{aligned}\frac{\partial f}{\partial \gamma} &= \frac{\partial f}{\partial y_i} \frac{\partial y_i}{\partial \gamma} \\ &= \sum_{i=1}^m \frac{\partial f}{\partial y_i} \hat{x}_i\end{aligned}$$
- $$\begin{aligned}\frac{\partial f}{\partial \beta} &= \frac{\partial f}{\partial y_i} \frac{\partial y_i}{\partial \beta} \\ &= \sum_{i=1}^m \frac{\partial f}{\partial y_i}\end{aligned}$$

c.

$$\begin{aligned}\frac{\partial f}{\partial \hat{x}_i} &= \frac{\partial f}{\partial y_i} \frac{\partial y_i}{\partial \hat{x}_i} \\ &= \frac{\partial f}{\partial y_i} \gamma\end{aligned}$$

d.

$$\begin{aligned}\frac{\partial f}{\partial \sigma^2} &= \frac{\partial f}{\partial y_i} \frac{\partial y_i}{\partial \hat{x}_i} \frac{\partial \hat{x}_i}{\partial \sigma^2} \\ &= \sum_{i=1}^m \frac{\partial f}{\partial y_i} \gamma \frac{-1}{2} \frac{(x_i - \mu)}{\sqrt{(\sigma^2 + \epsilon)^3}}\end{aligned}$$

e.

$$\begin{aligned}\frac{\partial f}{\partial \mu} &= \frac{\partial f}{\partial y_i} \frac{\partial y_i}{\partial \hat{x}_i} \frac{\partial \hat{x}_i}{\partial \mu} + \frac{\partial f}{\partial y_i} \frac{\partial y_i}{\partial \hat{x}_i} \frac{\partial \hat{x}_i}{\partial \sigma^2} \frac{\partial \sigma^2}{\partial \mu} \\ &= \sum_{i=1}^m \frac{\partial f}{\partial y_i} \gamma \frac{-1}{\sqrt{(\sigma^2 + \epsilon)}} \\ &\quad + \sum_{i=1}^m \frac{\partial f}{\partial y_i} \gamma \frac{-1}{2} \frac{(x_i - \mu)}{\sqrt{(\sigma^2 + \epsilon)^3}} \frac{-2}{m} \sum_{i=1}^m (x_i - \mu)\end{aligned}$$

f.

$$\begin{aligned}\frac{\partial f}{\partial x_i} &= \frac{\partial f}{\partial \hat{x}_i} \frac{\partial \hat{x}_i}{\partial x_i} + \frac{\partial f}{\partial \sigma^2} \frac{\partial \sigma^2}{\partial x_i} + \frac{\partial f}{\partial \mu} \frac{\partial \mu}{\partial x_i} \\ \frac{\partial f}{\partial x_i} &= \frac{\partial f}{\partial y_i} \gamma \frac{1}{\sqrt{(\sigma^2 + \epsilon)^3}} \\ &\quad + \frac{\partial f}{\partial y_i} \gamma \frac{-1}{2} \frac{(x_i - \mu)}{\sqrt{(\sigma^2 + \epsilon)^3}} \frac{2}{m} (x_i - \mu) \\ &\quad + \frac{\partial f}{\partial y_i} \gamma \frac{-1}{\sqrt{(\sigma^2 + \epsilon)^3}} \\ &\quad + \frac{\partial f}{\partial y_i} \gamma \frac{-1}{2} \frac{(x_i - \mu)}{\sqrt{(\sigma^2 + \epsilon)^3}} \frac{-2}{m} \sum_{i=1}^m (x_i - \mu) \frac{1}{m}\end{aligned}$$

4 Practical [50 pts]

TODO: Provide a report detailing your experiments, results and discussion in this section.