

作业 1

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1.1. Block One: Gradients of some basic layers (30 points)

- (i) Given a BatchNorm layer, please calculate the gradients of the output  $y_i = \text{BN}_{\gamma, \beta}(x_i)$  with respect to the parameters of  $\gamma, \beta$  shown in Figure 4. (10 points)
- (ii) Given a dropout layer, please calculate the gradients of **the output of a dropout layer** with respect to **the input of a dropout layer**. (10 points)
- (iii) Given a Softmax function, please calculate the gradients of **the output of a Softmax function** with respect to **the input of a Softmax function**. (10 points)

解. (i)  $\frac{\partial y_i}{\partial \gamma} = \hat{x}_i, \frac{\partial y_i}{\partial \beta} = 1$

- (ii) 设 dropout 层的输入为  $\mathcal{DJ}$  ( $n_{1a}$  维), 其中第  $i$  个值为  $\mathcal{DJ}_i$ , 设 dropout 层的输出为  $\mathcal{DO}$  ( $n_{1a}$  维), 其中第  $j$  个值为  $\mathcal{DO}_j$ , 则:

$$\frac{\partial \mathcal{DO}_j}{\partial \mathcal{DJ}_i} = \begin{cases} 0 & i \neq j \\ 0 & i = j \wedge r_i < p \\ 1/(1-p) & i = j \wedge r_i \geq p \end{cases}$$

- (iii) 设 Softmax 函数的输入为  $\mathcal{SJ}$  ( $n_{yb}$  维), 其中第  $i$  个值为  $\mathcal{SJ}_i$ , 设 Softmax 函数的输出为  $\mathcal{SO}$  ( $n_{yb}$  维), 其中第  $j$  个值为  $\mathcal{SO}_j$ , 则:

- 若  $i = j$ :

$$\begin{aligned} \frac{\partial \mathcal{SO}_j}{\partial \mathcal{SJ}_i} &= \frac{\partial \frac{e^{\mathcal{SJ}_i}}{\sum_k e^{\mathcal{SJ}_k}}}{\partial \mathcal{SJ}_i} = \frac{e^{\mathcal{SJ}_i} \cdot (\sum_k e^{\mathcal{SJ}_k}) - e^{\mathcal{SJ}_i} \cdot e^{\mathcal{SJ}_i}}{(\sum_k e^{\mathcal{SJ}_k})^2} \\ &= \mathcal{SO}_j \cdot (1 - \mathcal{SO}_j) \end{aligned}$$

- 若  $i \neq j$ :

$$\frac{\partial \mathcal{SO}_j}{\partial \mathcal{SJ}_i} = \frac{\partial \frac{e^{\mathcal{SJ}_j}}{\sum_k e^{\mathcal{SJ}_k}}}{\partial \mathcal{SJ}_i} = \frac{-e^{\mathcal{SJ}_j} \cdot e^{\mathcal{SJ}_i}}{(\sum_k e^{\mathcal{SJ}_k})^2} = -\mathcal{SO}_i \cdot \mathcal{SO}_j$$

1.2. **Block Two: Feed-forward and back-propagation of the multi-task network (30 points)**

- (i) Finish the detailed **feed-forward computations** of a batch samples  $(\mathbf{x}, y_a, y_b)$  during a training iteration, coming with final predictions  $(\hat{y}_a, \hat{y}_b)$  of Task A, Task B. **(10 points)**
- (ii) Use the back-propagation algorithm we have learned in class and give **the gradients of the overall loss function with respect to the parameters at each layer** corresponding to a batch of samples. **(20 points)**

解.