

# Large Scale Machine Learning and Deep Learning

## Review Questions 2

1. What is cross validation and how does it work?

**Answer:** the regression slides, pages 73–75

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2. Mathematically show that the softmax function with two classes ( $k = 2$ ) is equivalent to the sigmoid function?

**Answer:** let's look at the multiclass logistic regression, with  $k = 2$ :

$$p(y^{(i)} = 0) = \frac{e^{\mathbf{w}_0^T \mathbf{x}^{(i)}}}{e^{\mathbf{w}_0^T \mathbf{x}^{(i)}} + e^{\mathbf{w}_1^T \mathbf{x}^{(i)}}} = \frac{e^{(\mathbf{w}_0 - \mathbf{w}_1)^T \mathbf{x}^{(i)}}}{1 + e^{(\mathbf{w}_0 - \mathbf{w}_1)^T \mathbf{x}^{(i)}}} = \frac{e^{-\mathbf{w}^T \mathbf{x}^{(i)}}}{1 + e^{-\mathbf{w}^T \mathbf{x}^{(i)}}}$$

where  $\mathbf{w} = -(\mathbf{w}_0 - \mathbf{w}_1)$ , then we have:

$$p(y^{(i)} = 1) = 1 - p(y^{(i)} = 0) = 1 - \frac{e^{-\mathbf{w}^T \mathbf{x}^{(i)}}}{1 + e^{-\mathbf{w}^T \mathbf{x}^{(i)}}} = \frac{1}{1 + e^{-\mathbf{w}^T \mathbf{x}^{(i)}}}$$

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3. As you know, in binomial logistic regression the **cost** between the true value  $y$  and the predicted value  $\hat{y}$  is measured as below:

$$\text{cost}(\hat{y}, y) = \begin{cases} -\log(\hat{y}) & \text{if } y = 1 \\ -\log(1 - \hat{y}) & \text{if } y = 0 \end{cases}$$

Explain why  $-\log$  is a proper function to compute the cost in logistic regression?

**Answer:** the classification slides, pages 17–18

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4. How are logistic regression cost, cross-entropy, and negative log-likelihood related?

**Answer:** the classification slides, pages 30–36

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5. Explain how a ROC curve works?

**Answer:** the classification slides, pages 63–64