

1

Multiple Choice 1 point

Suppose we are using squared loss $l(y, \hat{y}) = (y - \hat{y})^2$ in a regression problem with a linear prediction model: $\hat{y} = \mathbf{w}^T \mathbf{x}$.

What is the derivative of the absolute loss w.r.t. \mathbf{w} for a single example with feature vector \mathbf{x} and true label y ?

Note: Because the absolute loss is not differentiable at zero, you can assume that y is not equal to \hat{y} in this problem. Note that the sign function for a real valued input x (where x is not zero) is defined as $\text{sign}(x) = +1$ for $x > 0$ and $\text{sign}(x) = -1$ for $x < 0$

- ☒ $\mathbf{x} (\mathbf{w}^T \mathbf{x} - y)$
- ☐ $\mathbf{x} (y - \text{softmax}(\mathbf{w}^T \mathbf{x}))$
- ☐ $\mathbf{x} (\text{softmax}(\mathbf{w}^T \mathbf{x}) - y)$
- ☐ $\mathbf{x} (y - \mathbf{w}^T \mathbf{x})$

2

Multiple Choice 1 point

Suppose in a classification problem, the true label \mathbf{y} and the predicted label $\hat{\mathbf{y}}$ are both hard labels. Which of the following correctly describes the nature of the cross-entropy loss in this special case?

- ☐ It is zero
- ☐ It is +infinity
- ☒ It is 0 if \mathbf{y} is equal to $\hat{\mathbf{y}}$ and +infinity otherwise
- ☐ It is 0 if \mathbf{y} is equal to $\hat{\mathbf{y}}$ and 1 otherwise

3

Multiple Choice 1 point

Recall that cross-entropy loss $l(\mathbf{y}, \mathbf{\hat{y}})$ is defined as the sum of $-y_j \log(\hat{y}_j)$ over j . What is the partial derivative of cross-entropy loss w.r.t. \hat{y}_j ?

- ☒ $-y_j / \hat{y}_j$
- ☐ $\text{softmax}(\mathbf{\hat{y}})_j - y_j$
- ☐ $-\hat{y}_j / y_j$
- ☐ $\text{softmax}(\mathbf{y})_j - \hat{y}_j$

4

Multiple Choice 1 point

Suppose $\mathbf{\hat{y}} = \text{softmax}(\mathbf{o})$ and I send o_1 , the first component of \mathbf{o} , off to **+infinity** (plus infinity) while keeping other components unchanged. What happens to \hat{y}_1 , the first component of $\mathbf{\hat{y}}$?

- ☐ It goes to -infinity
- ☐ It goes to +infinity
- ☒ It goes to 1
- ☐ It goes to 0

5

Multiple Choice 1 point

Suppose $\mathbf{\hat{y}} = \text{softmax}(\mathbf{o})$ and I send o_1 , the first component of \mathbf{o} , off to **-infinity** (minus infinity) while keeping other components unchanged. What happens to \hat{y}_1 , the first component of $\mathbf{\hat{y}}$?

- ☐ It goes to -infinity
- ☐ It goes to +infinity
- ☒ It goes to 0
- ☐ It goes to 1

6

Multiple Choice 1 point

Suppose I have a uniform distribution on N outcomes. What is the entropy (in bits) of this distribution?

- ☒ $\log_2(N)$
- ☐ $\log_2(1/N)$
- ☐ $1/N \log_2(N)$
- ☐ $1/N \log_2(1/N)$

7

Multiple Choice 1 point

Which of the following is NOT true regarding the relationship between TensorFlow and Keras?

- ☒ TensorFlow offers a more a high level API (application programming interface) than Keras
- ☐ First release of Keras was before the first release of Tensorflow
- ☐ Keras was originally built on top of Theano, a precursor of TensorFlow
- ☐ Keras API occupies a front-and-center place in TensorFlow 2.0

8

Multiple Choice 1 point

What is a feature of TensorFlow that is not all supported by NumPy?

- ☒ Ability to numerically compute gradients of any differentiable expression
- ☐ Ability to create vectors and matrices
- ☐ Ability to manipulate vectors and matrices
- ☐ A vast library of predefined mathematical functions such a sqrt, log, exp, etc.

9

Multiple Choice 1 point

What does the "G" in "GPU" stand for?

- ☒ graphics
- ☐ grammar
- ☐ gradient
- ☐ google

10

Multiple Choice 1 point

The name "Jupyter" includes a references to 3 major programming languages for data science. Which of the following is one of them?

- ☒ R
- ☐ Java
- ☐ JavaScript
- ☐ Ruby