Congratulations! You passed!

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1/1 point

1.

$$\widehat{J(\overrightarrow{\mathbf{w}},b)} = \frac{1}{m} \sum_{i=1}^{m} L(f_{\overrightarrow{\mathbf{w}},b}(\overrightarrow{\mathbf{x}}^{(i)}), \mathbf{y}^{(i)})$$

In this lecture series, "cost" and "loss" have distinct meanings. Which one applies to a single training example?

Loss

⊘ Correct

In these lectures, loss is calculated on a single training example. It is worth noting that this definition is not universal. Other lecture series may have a different definition.

☐ Cost

2.

- ☐ Both Loss and Cost
- Neither Loss nor Cost

1 / 1 point

Simplified loss function

$$L(f_{\overrightarrow{w},b}(\overrightarrow{\mathbf{x}}^{(i)}), \mathbf{y}^{(i)}) = \begin{cases} -\log(f_{\overrightarrow{w},b}(\overrightarrow{\mathbf{x}}^{(i)})) & \text{if } \mathbf{y}^{(i)} = 1\\ -\log(1 - f_{\overrightarrow{w},b}(\overrightarrow{\mathbf{x}}^{(i)})) & \text{if } \mathbf{y}^{(i)} = 0 \end{cases}$$

$$L(f_{\overrightarrow{w},b}(\overrightarrow{\mathbf{x}}^{(i)}), \mathbf{y}^{(i)}) = -\mathbf{y}^{(i)}\log(f_{\overrightarrow{w},b}(\overrightarrow{\mathbf{x}}^{(i)})) - (1 - \mathbf{y}^{(i)})\log(1 - f_{\overrightarrow{w},b}(\overrightarrow{\mathbf{x}}^{(i)}))$$

For the simplified loss function, if the label $y^{\left(i\right)}=0$, then what does this expression simplify to?

- \bigcirc $-\log(1-f_{\vec{\mathbf{w}},b}(\mathbf{x}^{(i)}))$
- $\bigcirc -\log(1-f_{ec{\mathbf{w}},b}(\mathbf{x}^{(i)})) log(1-f_{ec{\mathbf{w}},b}(\mathbf{x}^{(i)}))$
- $\bigcirc \log(f_{\vec{w},b}(\mathbf{x}^{(i)}))$
- igcirc $\log(1-f_{ec{\mathbf{w}},b}(\mathbf{x}^{(i)})) + log(1-f_{ec{\mathbf{w}},b}(\mathbf{x}^{(i)}))$
 - \bigcirc Correct When $y^{(i)}=0$, the first term reduces to zero.