1 / 1 point

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

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

Loss

- 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.
- Both Loss and Cost
- Neither Loss nor Cost

1 / 1 point

Simplified loss function

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

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

- For the simplified loss function, if the label $y^{(i)} = 0$, then what does this expression simplify to?
 - $-\log(1-f_{\vec{\mathbf{w}},b}(\mathbf{x}^{(i)}))$

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