1 / 1 point

Gradient descent is an algorithm for finding values of parameters w and b that minimize the cost function J.

repeat until convergence {

$$w = w - \alpha \frac{\partial}{\partial w} J(w, b)$$
$$b = b - \alpha \frac{\partial}{\partial b} J(w, b)$$

When $\frac{\partial J(w,b)}{\partial w}$ is a negative number (less than zero), what happens to w after one update step?

- w decreases
- w increases.
- It is not possible to tell if w will increase or decrease.

Correct
The learning rate is always a positive number, so if you take W minus a negative number, you end up with a new value for W that is larger (more positive)

1 / 1 point

- For linear regression, what is the update step for parameter b?

 - $b = b \alpha \frac{1}{m} \sum_{i=1}^{m} (f_{w,b}(x^{(i)}) y^{(i)})$ $b = b \alpha \frac{1}{m} \sum_{i=1}^{m} (f_{w,b}(x^{(i)}) y^{(i)}) x^{(i)}$

Correct The update step is $b=b-\alpha\frac{\partial J(w,b)}{\partial w}$ where $\frac{\partial J(w,b)}{\partial b}$ can be computed with this expression: $\sum_{i=1}^m (f_{w,b}(x^{(i)})-y^{(i)})$