Congratulations! You passed!

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To pass 70% or higher

Go to next item

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

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

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

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

- left w increases.
- igcup It is not possible to tell if w will increase or decrease.
- $\bigcirc w$ stays the same
- igcup w decreases

⊘ 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?

$$left b = b - lpha rac{1}{m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)})$$

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