Congratulations! You passed!

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Go to next item

1.

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

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

1/1 point

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

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

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

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