## Congratulations! You passed!

Grade received~100% **Latest Submission**  $\textbf{Grade}\,100\%$ 

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  $\frac{\partial J(w,b)}{\partial w}$  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.
- $\bigcirc \ w$  stays the same
- $\bigcirc w$  decreases
- 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 positive).

2.

1/1 point

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

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

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

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