## Congratulations! You passed!

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Go to next item

1/1 point

- 1. Which of the following can address overfitting?
  - Select a subset of the more relevant features.
  - ✓ Correct

If the model trains on the more relevant features, and not on the less useful features, it may generalize

better to new examples.

Remove a random set of training examples

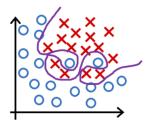
Regularization is used to reduce overfitting.

- Apply regularization
- Collect more training data

If the model trains on more data, it may generalize better to new examples

2. You fit logistic regression with polynomial features to a dataset, and your model looks like this.

1/1 point



What would you conclude? (Pick one)

- The model has high bias (underfit). Thus, adding data is likely to help
- O The model has high bias (underfit). Thus, adding data is, by itself, unlikely to help much.
- The model has high variance (overfit). Thus, adding data is, by itself, unlikely to help much.
- The model has high variance (overfit). Thus, adding data is likely to help

 $The \ model \ has \ high \ variance \ (it \ over fits \ the \ training \ data). \ Adding \ data \ (more \ training \ examples) \ can \ help.$ 

## Regularization

1/1 point

regularization term  $\min_{\vec{w},b} J(\vec{w},b) = \min_{\vec{w},b} \left( \frac{1}{2m} \sum_{i=1}^{m} (f_{\vec{w},b}(\vec{x}^{(i)}) - y^{(i)})^2 + \frac{\lambda}{2m} \sum_{j=1}^{n} w_j^2 \right)$ 

Suppose you have a regularized linear regression model. If you increase the regularization parameter  $\lambda$ , what do you expect to happen to the parameters  $w_1, w_2, ..., w_n$ ?

- lacksquare This will reduce the size of the parameters  $w_1, w_2, ..., w_n$
- igcup This will increase the size of the parameters  $w_1,w_2,...,w_n$

Regularization reduces overfitting by reducing the size of the parameters  $w_1, w_2, \dots w_n$ .