## SupervisedMachineLearning-WEEK-3-QUIZ-4

Link: <u>SupervisedMachineLearning-WEEK-3-QUIZ-4</u>

## Your grade: 100%

Your latest: 100% • Your highest: 100% • To pass you need at least 80%. We keep your highest score.

Next item ightarrow

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.

- Apply regularization
- **⊘** Correct

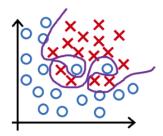
Regularization is used to reduce overfitting.

- Collect more training data
- **⊘** Correct

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

- ☐ Remove a random set of training 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)

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

The model has high variance (it overfits the training data). Adding data (more training examples) can help.

Regularization

1/1 point

 $\min_{\vec{w},b} J(\vec{w},b) = \min_{\vec{w},b} \left( \frac{1}{2m} \sum_{i=1}^{m} (f_{\vec{w},b}(\vec{x}^{(i)}) - y^{(i)})^2 + \frac{\lambda}{2m} \sum_{j=1}^{n} w_j^2 \right)$ uppose you have a residue of the second suppose of the second suppos

- 3. Suppose you have a regularized linear regression model. If you increase the regularization parameter  $\lambda$ , what do you expect to happen to the parameters  $w_1, w_2, ..., w_n$ ?
  - $\bigcirc$  This will increase the size of the parameters  $w_1, w_2, ..., w_n$
  - lacksquare This will reduce the size of the parameters  $w_1, w_2, ..., w_n$

**⊘** Correct

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