

Congratulations! You passed!

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 Grade received **100%** Latest Submission Grade **100%** To pass 80% or higher

1. Which of the following can address overfitting?

1 / 1 point

- ☐ Remove a random set of training examples
- ☒ 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.

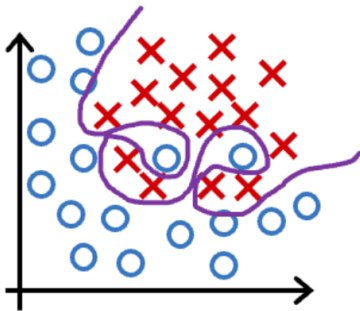
- ☒ Collect more training data

✓ **Correct**

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

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 variance (overfit). Thus, adding data is likely to help
- ☐ 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 bias (underfit). Thus, adding data is likely to help

✓ **Correct**

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

 3.

Regularization

1 / 1 point

$$\min_{\vec{w}} J(\vec{w}; h) = \min_{\vec{w}} \left[\underbrace{\frac{1}{m} \sum_{i=1}^m (f_{\vec{w}}(\vec{x}^{(i)}) - y^{(i)})^2}_{\text{mean squared error}} + \underbrace{\frac{\lambda}{2} \sum_{j=1}^n w_j^2}_{\text{regularization term}} \right]$$

$$\vec{w}, b = \arg \min_{\vec{w}, b} \left[2m \sum_{i=1}^m \ell(w, b; x^{(i)}, y^{(i)}) + \lambda \sum_{j=1}^n w_j^2 \right]$$

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

- ☒ This will reduce the size of the parameters w_1, w_2, \dots, w_n
- ☐ This will increase the size of the parameters w_1, w_2, \dots, w_n

✓ **Correct**

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