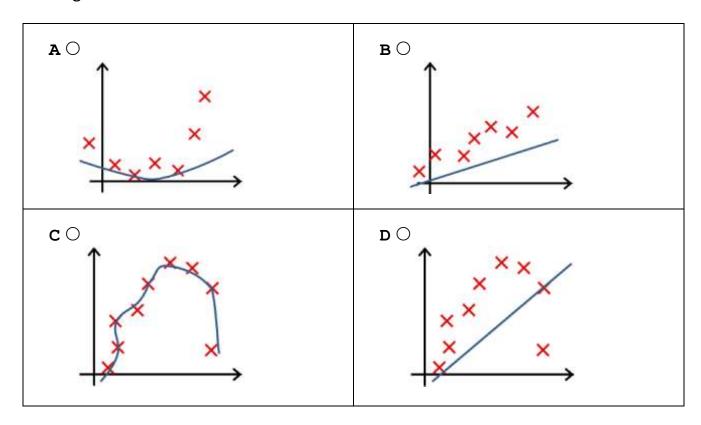
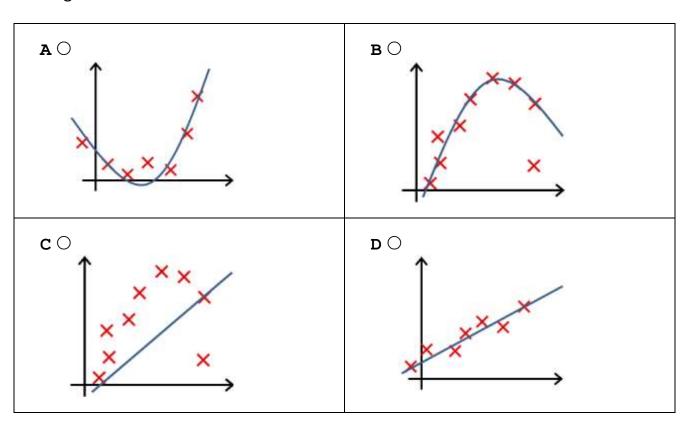
# Question 1

In which one of the following figures do you think the hypothesis has overfit the training set?



#### **Question 2**

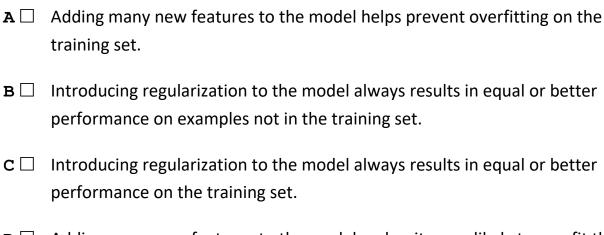
In which one of the following figures do you think the hypothesis has underfit the training set?



#### **Question 3**

You are training a classification model with logistic regression. Which of the following statements are true? Check all that apply.

Which of the following are true? Check all that apply.



D \( \text{Adding many new features to the model makes it more likely to overfit the training set.} \)

### **Question 4**

Suppose you ran logistic regression twice, once with  $\lambda$  = 0, and once with  $\lambda$  = 1. One of the times, you got parameters  $\theta = \begin{bmatrix} 26.29 \\ 65.41 \end{bmatrix}$ , and the other time you got  $\theta = \begin{bmatrix} 2.75 \\ 1.32 \end{bmatrix}$ . However, you forgot which value of  $\lambda$  corresponds to which value of  $\theta$ . Which one do you think corresponds to  $\lambda$  = 1?

$$\mathbf{A} \bigcirc \quad \theta = \begin{bmatrix} 26.29 \\ 65.41 \end{bmatrix}$$

$$\mathbf{B} \bigcirc \quad \theta = \begin{bmatrix} 2.75 \\ 1.32 \end{bmatrix}$$

## **Question 5**

Which of the following statements about regularization are true? Check all that apply.

- **A**  $\square$  Using a very large value of  $\lambda$  cannot hurt the performance of your hypothesis; the only reason we do not set  $\lambda$  to be too large is to avoid numerical problems.
- **B**  $\square$  Using too large a value of  $\lambda$  can cause your hypothesis to overfit the data; this can be avoided by reducing  $\lambda$ .
- Because regularization causes  $J(\theta)$  to no longer be convex, gradient descent may not always converge to the global minimum (when  $\lambda > 0$ , and when using an appropriate learning rate  $\alpha$ ).
- **D**  $\square$  Using too large a value of  $\lambda$  can cause your hypothesis to underfit the data.