

# Regularization

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## Question 1

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You are training a classification model with logistic regression.

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

- ☐ Adding many new features to the model helps prevent overfitting on the training set.
  - ☐ Introducing regularization to the model always results in equal or better performance on the training set.
  - ☐ Introducing regularization to the model always results in equal or better performance on examples not in the training set.
  - ☒ **Adding a new feature to the model always results in equal or better performance on the training set.**
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## Question 2

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Suppose you ran logistic regression twice, once with  $\lambda = 0$ , and once with  $\lambda = 1$ .

One of the times, you got parameters  $\theta = \begin{bmatrix} 81.47 \\ 12.69 \end{bmatrix}$ , and the other time you got  $\theta = \begin{bmatrix} 13.01 \\ 0.91 \end{bmatrix}$ .

However, you forgot which value of  $\lambda$  corresponds to which value of  $\theta$ .

Which one do you think corresponds to  $\lambda = 1$ ?

- ☒  $\theta = \begin{bmatrix} 13.01 \\ 0.91 \end{bmatrix}$
  - ☐  $\theta = \begin{bmatrix} 81.47 \\ 12.69 \end{bmatrix}$
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## Question 3

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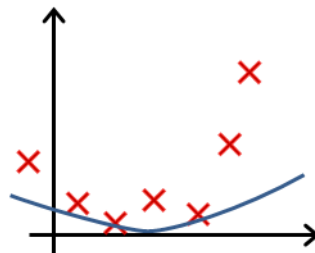
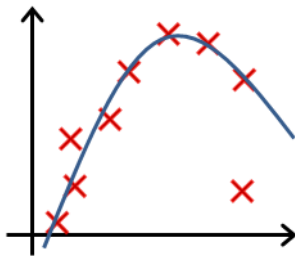
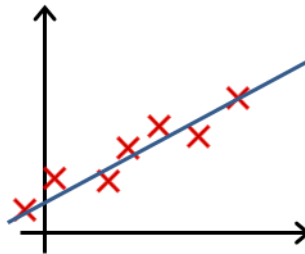
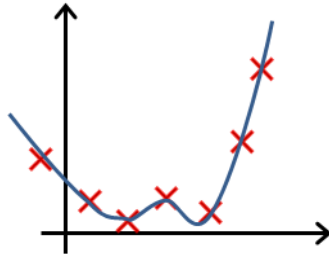
Which of the following statements about regularization are true? Check all that apply.

- ☐ Because logistic regression outputs values  $0 \leq h_{\theta}(x) \leq 1$ , its range of output values can only be "shrunk" slightly by regularization anyway, so regularization is generally not helpful for it.
- ☒ **Using too large a value of  $\lambda$  can cause your hypothesis to underfit the data.**
- ☐ 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  $\alpha$ ).

- ☐ 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.

## Question 4

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



## Question 5

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

