

# Week 3

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## Quiz 2-Regularization

**Ques1.** You are training a classification model with logistic regression. Which of the following statements are true? Check all that apply.

- A) Adding many new features to the model helps prevent overfitting on the training set.
- B) Adding a new feature to the model always results in equal or better performance on the training set.
- C) Introducing regularization to the model always results in equal or better performance on the training set.
- D) Introducing regularization to the model always results in equal or better performance on examples not in the training set.

**Answer:**

- A) Wrong. This will lead to overfitting
- B) Correct.
- C) Wrong. Can lead to underfitting
- D) Wrong. Underfitting will lead to worse performance on egs not in the training set.

## Ques2.

Suppose you ran logistic regression twice, once with  $\lambda = 0$ , and once with  $\lambda = 1$ . One of the times, you got

parameters  $\theta = \begin{bmatrix} 74.81 \\ 45.05 \end{bmatrix}$ , and the other time you got

$\theta = \begin{bmatrix} 1.37 \\ 0.51 \end{bmatrix}$ . However, you forgot which value of

$\lambda$  corresponds to which value of  $\theta$ . Which one do you

think corresponds to  $\lambda = 1$ ?

☒  $\theta = \begin{bmatrix} 1.37 \\ 0.51 \end{bmatrix}$

☐  $\theta = \begin{bmatrix} 74.81 \\ 45.05 \end{bmatrix}$

## Answer:

As when  $\lambda = 1$ , we add the regularization term which will penalize when  $\theta$  is big. Thus, when  $\lambda = 1$ ,  $\theta$  will be relatively smaller than without regularization.

**Ques3.** Which of the following statements about regularization are true? Check all that apply.

A) Consider a classification problem. Adding regularization may cause your classifier to incorrectly classify some training examples (which it had correctly classified when not using regularization, i.e. when  $\lambda=0$ ).

B) 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.

C) 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.

D) Using too large a value of  $\lambda$  can cause your hypothesis to overfit the data; this can be avoided by reducing  $\lambda$ .

**Answer:**

A) Correct.

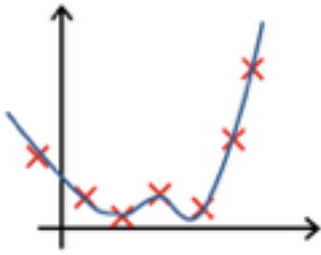
B) Wrong.

C) Wrong. Very large  $\lambda$  can lead to underfitting problem.

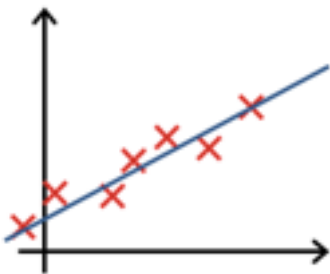
D) Wrong. Very large  $\lambda$  can lead to underfitting problem.

**Ques4.** In Which one of the following figures do you think the hypothesis has overfit the training set?

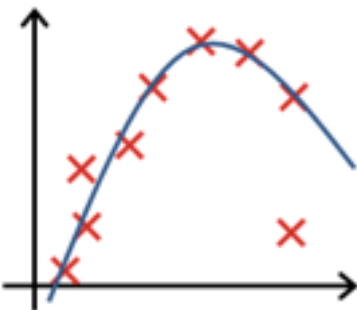
☒ Figure:



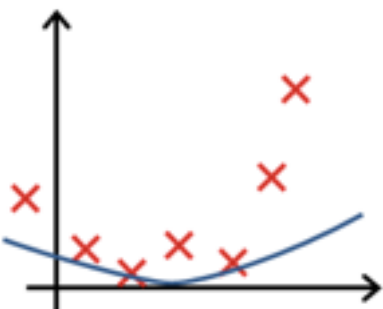
☐ Figure:



☐ Figure:



☐ Figure:



**Ques5.** In Which one of the following figures do you think the hypothesis has underfit the training set?



Figure:

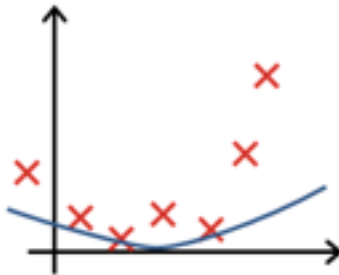


Figure:

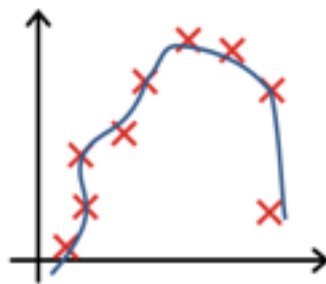


Figure:

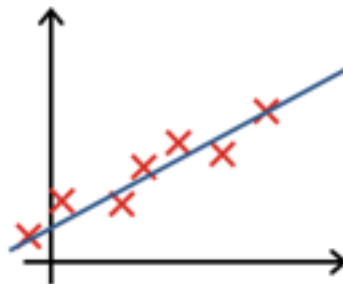


Figure:

