

## Regularization

TOTAL POINTS 5

1.	You are	training	а	classification	model	with	logistic

1 point

regression. Which of the following statements are true? Check

all that apply.

Adding a new feature to the model always results in equal or better performance on the training set.

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

Introducing regularization to the model always results in equal or better performance on examples not in the training set.

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

parameters  $heta = \begin{bmatrix} 26.29 \\ 65.41 \end{bmatrix}$  , and the other time you got



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

think corresponds to  $\lambda=1$ ?

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

$$\bullet = \begin{bmatrix} 2.75 \\ 1.32 \end{bmatrix}$$

1 point



3. Which of the following statements about regularization are

1 point

true? Check all that apply.



reducing  $\lambda$ .

Because logistic regression outputs values  $0 \le h_{\theta}(x) \le 1$ , its range of output values can only be "shrunk" slightly by regularization anyway, so regularization is generally not helpful for it.

ightharpoonup 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  $\lambda=0$ ).

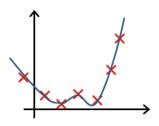
 $\$  Using a very large value of  $\lambda$  cannot hurt the performance of your hypothesis; the only reason we do not set  $\lambda$  to be too large is to avoid numerical problems.



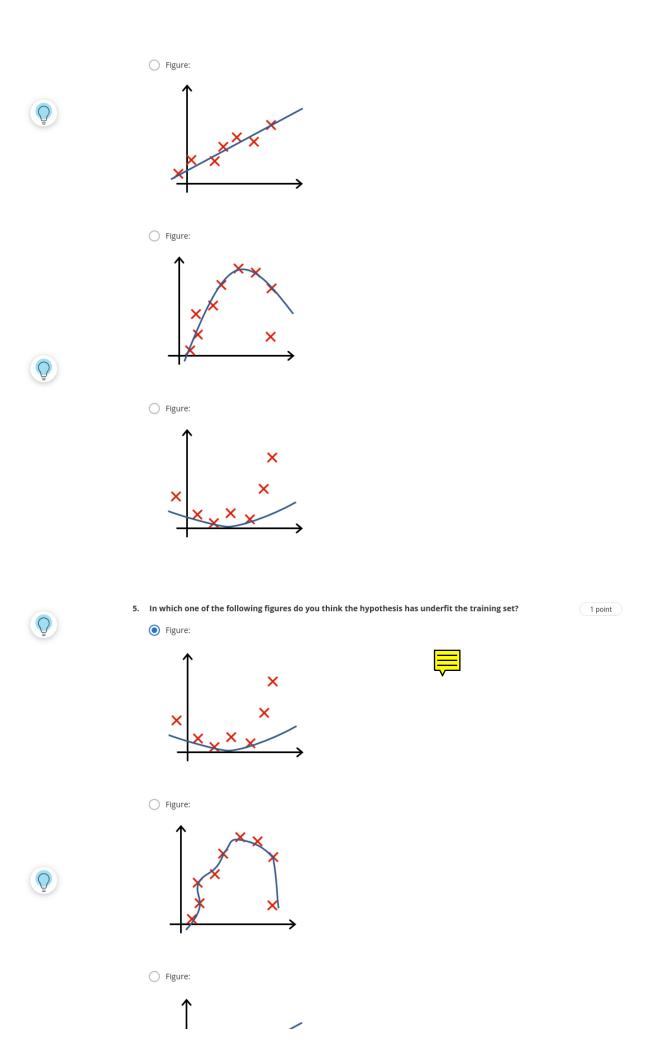
I. In which one of the following figures do you think the hypothesis has overfit the training set?

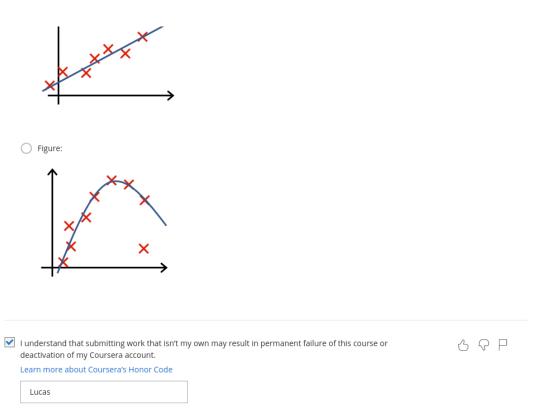
1 point

Figure:









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