

✔ Congratulations! You passed!

Grade received 100% To pass 80% or higher

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## Logistic Regression

Total points 10

1. When performing logistic regression on sentiment analysis, you represented each tweet as a vector of ones and zeros. However your model did not work well. Your training cost was reasonable, but your testing cost was just not acceptable. What could be a possible reason?

1 / 1 point

- ☒ The vector representations are sparse and therefore it is much harder for your model to learn anything that could generalize well to the test set.
- ☐ You probably need to increase your vocabulary size because it seems like you have very little features.
- ☐ Logistic regression does not work for sentiment analysis, and therefore you should be looking at other models.
- ☐ Sparse representations require a good amount of training time so you should train your model for longer

✔ Correct  
This is correct.

2. Which of the following are examples of text preprocessing?

1 / 1 point

- ☒ Stemming, or the process of reducing a word to its word stem.

✔ Correct  
This is correct.

- ☒ Lowercasing, which is the process of removing changing all capital letter to lower case.

✔ Correct  
This is correct.

- ☒ Removing stopwords, punctuation, handles and URLs

✔ Correct  
This is correct.

- ☐ Adding new words to make sure all the sentences make sense

3. The sigmoid function is defined as  $h(x^{(i)}, \theta) = \frac{1}{1 + e^{-\theta^T x^{(i)}}}$ . Which of the following is true.

1 / 1 point

- ☐ Large positive values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  closer to 1 and large negative values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  close to -1.
- ☒ Large positive values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  closer to 1 and large negative values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  close to 0.
- ☐ Small positive values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  closer to 1 and large positive values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  close to 0.
- ☐ Small positive values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  closer to 0 and large negative values of  $\theta^T x^{(i)}$  will make  $h(x^{(i)}, \theta)$  close to -1.

✓ **Correct**  
This is correct.

4. The cost function for logistic regression is defined as  $J(\theta) = -\frac{1}{m} \sum_{i=1}^m [y^{(i)} \log h(x^{(i)}, \theta) + (1 - y^{(i)}) \log (1 - h(x^{(i)}, \theta))]$ . Which of the following is true about the cost function above. Mark all the correct ones.

1 / 1 point

✓ When  $y^{(i)} = 1$ , as  $h(x^{(i)}, \theta)$  goes close to 0, the cost function approaches  $\infty$ .

✓ **Correct**  
This is correct.

□ When  $y^{(i)} = 1$ , as  $h(x^{(i)}, \theta)$  goes close to 0, the cost function approaches 0.

✓ When  $y^{(i)} = 0$ , as  $h(x^{(i)}, \theta)$  goes close to 0, the cost function approaches 0.

✓ **Correct**  
This is correct.

□ When  $y^{(i)} = 0$ , as  $h(x^{(i)}, \theta)$  goes close to 0, the cost function approaches  $\infty$ .

5. For what value of  $\theta^T x$  in the sigmoid function does  $h(x^{(i)}, \theta) = 0.5$ .

1 / 1 point

0

✓ **Correct**

6. Select all that apply. When performing logistic regression for sentiment analysis using the method taught in this week's lecture, you have to:

1 / 1 point

✓ Performing data processing.

✓ **Correct**  
This is correct.

✓ Create a dictionary that maps the word and the class that word is found in to the number of times that word is found in the class.

✓ **Correct**  
This is correct.

□ Create a dictionary that maps the word and the class that word is found in to see if that word shows up in the class.

✓ For each tweet, you have to create a **positive feature** with the sum of positive counts of each word in that tweet. You also have to create a **negative feature** with the sum of negative counts of each word in that tweet.

✓ **Correct**  
This is correct.

7. When training logistic regression, you have to perform the following operations in the desired order.

1 / 1 point

○ Initialize parameters, get gradient, classify/predict, update, get loss, repeat

○ Initialize parameters, classify/predict, get gradient, update, get loss, repeat

☒ Initialize parameters, classify/predict, get gradient, update, get loss, repeat

☐ Initialize parameters, get gradient, update, classify/predict, get loss, repeat

☐ Initialize parameters, get gradient, update, get loss, classify/predict, repeat

✓ **Correct**  
This is correct.

8. Assuming we got the classification correct, where  $y^{(i)} = 1$  for some specific example  $i$ . This means that  $h(x^{(i)}, \theta) > 0.5$ . Which of the following has to hold:

1 / 1 point

☐ Our prediction,  $h(x^{(i)}, \theta)$  for this specific training example is exactly equal to its corresponding label  $y^{(i)}$ .

☐ Our prediction,  $h(x^{(i)}, \theta)$  for this specific training example is less than  $(1 - y^{(i)})$ .

☐ Our prediction,  $h(x^{(i)}, \theta)$  for this specific training example is less than  $(1 - h(x^{(i)}, \theta))$ .

☒ Our prediction,  $h(x^{(i)}, \theta)$  for this specific training example is greater than  $(1 - h(x^{(i)}, \theta))$ .

✓ **Correct**  
This is correct.

9. What is the purpose of gradient descent? Select all that apply.

1 / 1 point

☒ Gradient descent allows us to learn the parameters  $\theta$  in logistic regression as to minimize the loss function  $J$ .

✓ **Correct**  
This is correct.

☐ Gradient descent allows us to learn the parameters  $\theta$  in logistic regression as to maximize the loss function  $J$ .

☒ Gradient descent,  $grad\_theta$  allows us to update the parameters  $\theta$  by computing  $\theta = \theta - \alpha * grad\_theta$

✓ **Correct**  
This is correct.

☐ Gradient descent,  $grad\_theta$  allows us to update the parameters  $\theta$  by computing  $\theta = \theta + \alpha * grad\_theta$

10. What is a good metric that allows you to decide when to stop training/trying to get a good model? Select all that apply.

1 / 1 point

☒ When your accuracy is good enough on the test set.

✓ **Correct**  
This is correct.

☐ When your accuracy is good enough on the train set.

☒ When you plot the cost versus (# of iterations) and you see that your the loss is converging (i.e. no longer changes as much).

✓ **Correct**  
This is correct.

☐ When you plot the cost versus (# of iterations) and you see that your the loss is converging (i.e. no longer changes as much).

□ when  $\alpha$ , your step size is neither too small nor too large.