

1. If you have 10,000,000 examples, how would you split the train/dev/test set?

☒ 98% train . 1% dev . 1% test

Correct

☐ 60% train . 20% dev . 20% test

☐ 33% train . 33% dev . 33% test

2. The dev and test set should:

☒ Come from the same distribution

Correct

☐ Come from different distributions

☐ Be identical to each other (same (x,y) pairs)

☐ **Have the same number of examples**

3. If your Neural Network model seems to have high variance, what of the following would be promising things to try?

☒ Add regularization

Correct

☐ Make the Neural Network deeper

Un-selected is correct

☐ Increase the number of units in each hidden layer

Un-selected is correct

☒ Get more training data

Correct

☐ Get more test data

Un-selected is correct

4. You are working on an automated check-out kiosk for a supermarket, and are building a classifier for apples, bananas and oranges. Suppose your classifier obtains a training set error of 0.5%, and a dev set error of 7%. Which of the following are promising things to try to improve your classifier? (Check all that apply.)

☒ Increase the regularization parameter  $\lambda$

Correct

☐ Decrease the regularization parameter  $\lambda$

Un-selected is correct

☒ Get more training data

Correct

☐ Use a bigger neural network

Un-selected is correct

5. What is weight decay?

- ☐ The process of gradually decreasing the learning rate during training.
- ☐ A technique to avoid vanishing gradient by imposing a ceiling on the values of the weights.
- ☐ Gradual corruption of the weights in the neural network if it is trained on noisy data.
- ☒ A regularization technique (such as L2 regularization) that results in gradient descent shrinking the weights on every iteration.

Correct

6. What happens when you increase the regularization hyperparameter  $\lambda$ ?

- ☒ Weights are pushed toward becoming smaller (closer to 0)

Correct

- ☐ Weights are pushed toward becoming bigger (further from 0)
- ☐ Doubling  $\lambda$  should roughly result in doubling the weights
- ☐ Gradient descent taking bigger steps with each iteration (proportional to  $\lambda$ )

---

7. With the inverted dropout technique, at test time:

- ☒ You do not apply dropout (do not randomly eliminate units) and do not keep the  $1/\text{keep\_prob}$  factor in the calculations used in training

Correct

- ☐ You apply dropout (randomly eliminating units) but keep the  $1/\text{keep\_prob}$  factor in the calculations used in training.
- ☐ You apply dropout (randomly eliminating units) and do not keep the  $1/\text{keep\_prob}$  factor in the calculations used in training
- ☐ You do not apply dropout (do not randomly eliminate units), but keep the  $1/\text{keep\_prob}$  factor in the calculations used in training.

8. Increasing the parameter `keep_prob` from (say) 0.5 to 0.6 will likely cause the following:  
(Check the two that apply)

☐ Increasing the regularization effect

Un-selected is correct

☒ Reducing the regularization effect

Correct

☐ Causing the neural network to end up with a higher training set error

Un-selected is correct

☒ Causing the neural network to end up with a lower training set error

Correct

9. Which of these techniques are useful for reducing variance (reducing overfitting)? (Check all that apply.)

☐ Exploding gradient

Un-selected is correct

☒ Dropout

Correct

☒ L2 regularization

Correct

☐ Gradient Checking

Un-selected is correct

☐ Vanishing gradient

Un-selected is correct

10. Why do we normalize the inputs  $x$ ?

☐ It makes it easier to visualize the data

☒ It makes the cost function faster to optimize

Correct

☐ Normalization is another word for regularization--It helps to reduce variance

☐ It makes the parameter initialization faster