

# Practical aspects of deep learning

LATEST SUBMISSION GRADE

100%

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

1 / 1 point

- ☐ 33% train . 33% dev . 33% test
- ☒ 98% train . 1% dev . 1% test
- ☐ 60% train . 20% dev . 20% test



Correct

2. The dev and test set should:

1 / 1 point

- ☒ Come from the same distribution
- ☐ Come from different distributions
- ☐ Be identical to each other (same (x,y) pairs)
- ☐ Have the same number of examples



Correct

3. If your Neural Network model seems to have high bias, what of the following would be promising things to try? (Check all that apply.)

1 / 1 point

- ☒ Make the Neural Network deeper



Correct

☒ Increase the number of units in each hidden layer



**Correct**

☐ Add regularization

☐ Get more test data

☐ Get more training data

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.)

**1 / 1 point**

☒ Increase the regularization parameter  $\lambda$



**Correct**

☐ Decrease the regularization parameter  $\lambda$

☒ Get more training data



**Correct**

☐ Use a bigger neural network

5. What is weight decay?

**1 / 1 point**

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

**Correct**



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

1 / 1 point

- ☒ Weights are pushed toward becoming smaller (closer to 0)
- ☐ 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  $\lambda$ )



Correct

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

1 / 1 point

- ☒ 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
- ☐ 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.



Correct

8. Increasing the parameter  $\text{keep\_prob}$  from (say) 0.5 to 0.6 will likely cause the following:  
(Check the two that apply)

1 / 1 point

- ☐ Increasing the regularization effect
- ☒ Reducing the regularization effect



Correct

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

- ☒ 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.) **1 / 1 point**

- ☒ Dropout

✓ **Correct**

- ☐ Vanishing gradient

- ☒ L2 regularization

✓ **Correct**

- ☐ Gradient Checking

- ☒ Data augmentation

✓ **Correct**

- ☐ Exploding gradient

- ☐ Xavier initialization

10. Why do we normalize the inputs  $x$ ? **1 / 1 point**

- ☐ It makes it easier to visualize the data
- ☐ It makes the parameter initialization faster
- ☐ Normalization is another word for regularization--It helps to reduce variance
- ☒ It makes the cost function faster to optimize



**Correct**