

# Practical aspects of deep learning

**10/10 points (100%)**

Quiz, 10 questions

 **Congratulations! You passed!**[Next Item](#)1 / 1  
points

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

1 / 1  
points

2.

The dev and test set should:



Come from the same distribution

**Correct**

Come from different distributions

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Have the same number of examples



1 / 1  
points

3.

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



Increase the number of units in each hidden layer



Un-selected is correct



Add regularization



Correct



Make the Neural Network deeper



Un-selected is correct



Get more test data



Un-selected is correct



Get more training data



Correct



1 / 1  
points

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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.)

10/10 points (100%)



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



1 / 1  
points

5.

What is weight decay?



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

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points

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  $\lambda$ )

1 / 1  
points

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

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points

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**1 / 1  
points

9.

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



L2 regularization



**Correct**

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Dropout

**Correct**

Exploding gradient

**Un-selected is correct**

Vanishing gradient

**Un-selected is correct**

Xavier initialization

**Un-selected is correct**

Gradient Checking

**Un-selected is correct**

Data augmentation

**Correct**1 / 1  
points

10.

Why do we normalize the inputs  $x$ ?

It makes the parameter initialization faster

