

# Practical aspects of deep learning

**6/10 points (60.00%)**

Quiz, 10 questions

## ✖ Try again once you are ready.

Required to pass: 80% or higher

You can retake this quiz up to 3 times every 8 hours.

[Back to Week 1](#)[Retake](#)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**

33% train . 33% dev . 33% test



60% train . 20% dev . 20% test

1 / 1  
points

2.

The dev and test set should:



Come from the same distribution

**Correct**

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

**Have the same number of examples**1 / 1  
points

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.)



Make the Neural Network deeper

**Correct**

Increase the number of units in each hidden layer

**Correct**

Get more test data

**Un-selected is correct**

Get more training data

**Un-selected is correct**

Add regularization

**Un-selected is correct**

0 / 1



points

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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  $\lambda$ **This should be selected**☐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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The process of gradually decreasing the learning rate during training.



1 / 1  
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  $\lambda$ )



0 / 1  
points

**7.**

With the inverted dropout technique, at test time:



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 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.

**This should not be selected**

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You are only randomly eliminating units) and do not keep the 1/keep\_prob factor in the calculations used in training

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1 / 1  
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**



0 / 1  
points

9.

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

☐

Exploding gradient

Un-selected is correct

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Xavier initialization



This should not be selected



Data augmentation



Correct



Vanishing gradient



Un-selected is correct



Dropout



Correct



L2 regularization



Correct



Gradient Checking



Un-selected is correct



0 / 1  
points

10.

Why do we normalize the inputs  $x$ ?



It makes the parameter initialization faster



It makes the cost function faster to optimize

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**This should not be selected**

☐ It makes it easier to visualize the data

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