## Congratulations! You passed!

If you have 10,000,000 examples, how would you split the train/dev/test set? 1. 98% train . 1% dev . 1% test Correct 60% train . 20% dev . 20% test 33% train . 33% dev . 33% test 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 If your Neural Network model seems to have high bias, what of the following would be promising things to try? (Check all that apply.) Get more test data **Un-selected is correct** Increase the number of units in each hidden layer **Correct** 

		Add regularization
		Un-selected is correct
		Make the Neural Network deeper
		Correct
		Get more training data
		Un-selected is correct
1/1	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.)
points		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?
1 / 1 points		A technique to avoid vanishing gradient by imposing a ceiling on the values of the weights.
		A regularization technique (such as L2 regularization) that results in gradient descent shrinking the weights on every iteration.
		Correct
		Gradual corruption of the weights in the neural network if it is trained on
		• • • • • • • • • • • • • • • • • • • •

		noisy data.
		The process of gradually decreasing the learning rate during training.
	6.	What happens when you increase the regularization hyperparameter lambda?
_		Weights are pushed toward becoming smaller (closer to 0)
1 / 1 points		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:
1/1		You do not apply dropout (do not randomly eliminate units), but keep the 1/keep_prob factor in the calculations used in training.
points		You apply dropout (randomly eliminating units) but keep the 1/keep_prob factor in the calculations used in training.
		You apply dropout (randomly eliminating units) and do not keep the 1/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/keep_prob factor in the calculations used in training
		Correct
	8.	Increasing the parameter keep_prob from (say) 0.5 to 0.6 will likely cause the following: (Check the two that apply)
1/1		Increasing the regularization effect
points		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.)  Exploding gradient
		Un-selected is correct
		Dropout
		Data augmentation  Correct
		L2 regularization  Correct
		Vanishing gradient Un-selected is correct
		Gradient Checking
		Un-selected is correct
		Xavier initialization
		Un-selected is correct
	10.	Why do we normalize the inputs $x$ ?
1/1		It makes the parameter initialization faster
points		It makes it easier to visualize the data  Normalization is another word for regularizationIt helps to reduce variance
		It makes the cost function factor to entimize

Correct