| 1. | If you have 10,000,000 examples, how would you split the train/dev/test set? |
|----|--|
| | 60% train . 20% dev . 20% test |
| | 98% train . 1% dev . 1% test |
| | Correct |
| | 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 |
| | |
| | |
| | |

| ٥. | promising things to try? (Check all that apply.) |
|----|---|
| 4. | Make the Neural Network deeper |
| | Correct |
| | Get more training data |
| | Un-selected is correct |
| | Add regularization |
| | Un-selected is correct |
| | Increase the number of units in each hidden layer |
| | Correct |
| | Get more test data |
| | Un-selected is correct |
| | 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? |
| | 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) |
| | 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 apply dropout (randomly eliminating units) but keep the 1/keep_prob factor in the calculations used in training. | |
| | | You do not apply dropout (do not randomly eliminate 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 | |
| | Corr | ect | |
| | | | |
| 8. | | sing the parameter keep_prob from (say) 0.5 to 0.6 will likely cause the following: the two that apply) | |
| | | Increasing the regularization effect | |
| | Un-s | elected is correct | |
| | | Reducing the regularization effect | |
| | Corr | ect | |
| | | Causing the neural network to end up with a higher training set error | |
| | Un-s | relected is correct | |
| | | | |
| | | Causing the neural network to end up with a lower training set error | |
| | This s | hould be selected | |
| | | | |

| 9. | Which of these techniques are useful for reducing variance (reducing overfitting)? (Check all that apply.) | | | |
|----|--|--|--|--|
| | Data augmentation | | | |
| | Correct | | | |
| | Gradient Checking | | | |
| | Un-selected is correct | | | |
| | Exploding gradient | | | |
| | Un-selected is correct | | | |
| | Vanishing gradient | | | |
| | Un-selected is correct | | | |
| | Xavier initialization | | | |
| | Un-selected is correct | | | |
| | L2 regularization | | | |
| | Correct | | | |
| | Dropout | | | |
| | Correct | | | |
| | | | | |

| 10. | Why do | Why do we normalize the inputs x ? | | |
|-----|--------|---|--|--|
| | | It makes the cost function faster to optimize | | |
| | Corre | ect | | |
| | | | | |
| | | It makes the parameter initialization faster | | |
| | | It makes it easier to visualize the data | | |
| | | Normalization is another word for regularizationIt helps to reduce variance | | |
| | | | | |