Congratulations! You passed!

Grade received 90%   **To pass** 80% or higher

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Practical aspects of deep learning

Latest Submission Grade 90%

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

○ 60% train . 20% dev . 20% test

○ 33% train . 33% dev . 33% test

◉ 98% train . 1% dev . 1% 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.)   **0 / 1 point**

☑ Add regularization

⊗ **This should not be selected**

☐ Make the Neural Network deeper

☐ Increase the number of units in each hidden layer

☑ Get more training data

⊗ **This should not be selected**

☐ Get more test 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**

◉ A regularization technique (such as L2 regularization) that results in gradient descent shrinking the weights on every iteration.

○ The process of gradually decreasing the learning rate during training.

○ 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.

⊘ **Correct**

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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**

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7. With the inverted dropout technique, at test time: **1 / 1 point**

○ 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

○ You do not apply dropout (do not randomly eliminate units), but keep the 1/keep_prob factor in the calculations used in training.

⊘ **Correct**

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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 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**

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9. Which of these techniques are useful for reducing variance (reducing overfitting)? (Check all that apply.)   **1 / 1 point**

☐ Vanishing gradient

☐ Xavier initialization

☐ Exploding gradient

☑ L2 regularization

⊘ **Correct**

☑ Dropout

⊘ **Correct**

☐ Gradient Checking

☑ Data augmentation

⊘ **Correct**

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

○ It makes it easier to visualize the data

○ Normalization is another word for regularization--It helps to reduce variance

○ It makes the parameter initialization faster

◉ It makes the cost function faster to optimize

⊘ **Correct**

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⊘ **Correct**