1.

## Congratulations! You passed!

Grade received 100% Latest Submission Grade 100% To pass 80% or higher

Go to next item

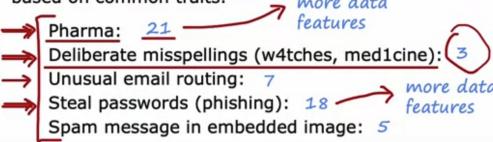
1/1 point

## Error analysis

 $m_{cv} = \frac{500}{500}$  examples in cross validation set.

Algorithm misclassifies 100 of them.

Manually examine 100 examples and categorize them based on common traits.



Which of these is a way to do error analysis?

- Manually examine a sample of the training examples that the model misclassified in order to identify common traits and trends.
- igcup Calculating the test error  $J_{test}$
- igcup Calculating the training error  $J_{train}$
- Ollecting additional training data in order to help the algorithm do better.

## 

Correct. By identifying similar types of errors, you can collect more data that are similar to these misclassified examples in order to train the model to improve on these types of examples.

Data augmentation

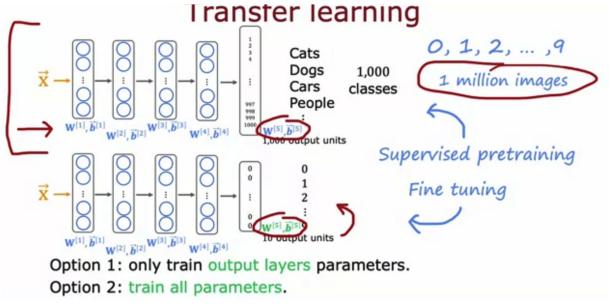
Augmentation: modifying an existing training example to create a new training example.

(X,Y)

1/1 point

<ul><li>•</li></ul>	Data augmentation
0	Error analysis
0	Machine learning diagnostic
0	Bias/variance analysis
(	Correct Yes! Modifying existing data (such as images, or audio) is called data augmentation.

3.



What are two possible ways to perform transfer learning? Hint: two of the four choices are correct.

✓ You can choose to train all parameters of the model, including the output layers, as well as the earlier layers.

**⊘** Correct

Correct. It may help to train all the layers of the model on your own training set. This may take more time compared to if you just trained the parameters of the output layers.

- Download a pre-trained model and use it for prediction without modifying or re-training it.
- You can choose to train just the output layers' parameters and leave the other parameters of the model fixed.
- ✓ Correct

 $Correct. The \ earlier \ layers \ of \ the \ model \ may \ be \ reusable \ as \ is, \ because \ they \ are \ identifying \ low \ level \ features \ that \ are \ relevant \ to \ your \ task.$ 

Given a dataset, pre-train and then further fine tune a neural network on the same dataset.

1/1 point