

## ✓ Congratulations! You passed!

Grade received **100%** To pass 75% or higher

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### Selecting and Training a Model

Latest Submission Grade 100%

1. Which of these is a more accurate description of a data-centric approach to ML development?

1 / 1 point

- ☒ Holding the neural network architecture fixed, work to improve the data to do well on the problem.
- ☐ Holding the training data fixed, work to improve your neural network's architecture to do well on the problem.

✓ **Correct**

That's right! Data-centric means you focus your efforts on improving the data to raise the system's performance, while keeping the code fixed.

2. Say you have an algorithm that diagnoses illnesses from medical X-rays, and achieves high average test set accuracy. What can you now say with high confidence about this algorithm? Check all that apply.

1 / 1 point

- ☐ It does well even on rare classes of diseases.
- ☐ Its diagnoses are roughly equally accurate on all genders and ethnicities, so we are confident it is not biased against any gender or ethnicity.
- ☐ The system can be safely deployed in a healthcare setting.
- ☒ None of the above.

✓ **Correct**

3. Which of these statements about establishing a baseline are accurate? Check all that apply.

1 / 1 point

- ☐ Open-source software should not be used to establish a baseline, since the performance of a good open source implementation might be too good and thus too hard to beat.
- ☒ For unstructured data problems, using human-level performance as the baseline can give an estimate of the irreducible error/Bayes error and what performance is reasonable to achieve.



**Correct**

That's right! For most unstructured data problems, human-level performance is a great estimate of Bayes error - an upper limit to your system's potential.

- ☒ Human level performance (HLP) is generally more effective for establishing a baseline on unstructured data problems (such as images and audio) than structured data problems



**Correct**

You're right! Humans perform well on unstructured data, like making sense of an image or a sound, but we aren't great at making sense of large amounts of structured data.

- ☒ It can be established based on an older ML system



**Correct**

That's right! You can establish a baseline using an older system or via a literature or open source search.

1 / 1 point

4. On a speech recognition problem, say you run the sanity-check test of trying to overfit a single training example. You pick a clearly articulated clip of someone saying "Today's weather", and the algorithm fails to fit even this single audio clip, and outputs "\_\_\_\_\_". What should you do?

- ☐ Train the algorithm on a larger dataset to help it to fit the data better.
- ☒ Debug the code/algorithm/hyperparameters to make it pass this sanity-check test first, before moving to larger datasets.
- ☐ Create a training set of this example repeated 100 times to force the algorithm to learn to fit this example well.
- ☐ Use data augmentation on this one audio clip to make sure the algorithm hears a variety of examples of "today's weather" to fit this phrase better.



**Correct**

That's right! Something is clearly wrong with the implementation if the algorithm is unable to overfit to a single training example! Find the root cause, fix the problem, and *then* move onto larger datasets.