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## Chapter 6: A Closer Look at Machines That Learn

### → Reading/Mining/Discussion Assignment

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Please ...

1. Read “Chapter 6: A Closer Look at Machines That Learn” of Melanie Mitchell’s “Artificial Intelligence: A Guide for Thinking Humans” book.
2. With respect to the questions presented for the “Chapter 6: A Closer Look at Machines That Learn” reading, construct a document containing question/answer pairs, one pair of each question, where the answers, with perhaps just a small number of exceptions, are more or less lifted from Melanie Mitchell’s text. Save your document as a **pdf** file.
3. Post your question/answer document to your web worksite.
4. Do your best to internalize your question/answer pairs in some sort of semantic sense, so that the answers are likely to come back to you when prompted by the questions.
5. Come to class for the discussion of “Chapter 6: A Closer Look at Machines That Learn,” when the time rolls around, prepared to participate in the discussion.
6. Please do all of this within one week of the “distribution” of this assignment.

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## Chapter 6: A Closer Look at Machines That Learn - Questions

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1. TRUE/FALSE - The learning-from-data approach of deep neural networks has generally proved to be more successful than the “good old-fashioned AI” strategy, in which human programmers construct explicit rules for intelligent behavior. However, contrary to what some media have reported, the learning process of ConvNets is not very humanlike.
2. Why does your professor like the previous question?
3. TRUE/FALSE - As we’ve seen, the most successful ConvNets learn via a *supervised-learning* procedure: they gradually change their weights as they process the examples in the training set again and again, over many epochs (that is, many passes through the training set), learning to classify each input as one of a fixed set of possible output categories.
4. List some significant differences between the way that humans learn about objects and the way that ConvNets learn about objects.
5. Why is it inaccurate to say that today’s successful ConvNets “learn on their own?”

6. In answer to the rhetorical question “Where does all of the data come from to fuel big data applications?,” MM answers “You - and probably everyone you know.” Please elaborate on the answer.
7. How do car companies acquire the big data (labelled images of pedestrians, cyclists and other obstacles) needed to train robo-cars?
8. What is the “long tail” phenomenon, and how does it relate to machines that learn (ConvNets)?
9. TRUE/FALSE - A commonly proposed solution to the long tail problem in AI systems is to complement supervised learning with unsupervised learning.
10. What is “unsupervised learning?”
11. What colorful remark did Yann LeCun make about unsupervised learning?
12. TRUE/FALSE - For general AI, almost all learning will have to be unsupervised, but no one has yet come up with the kinds of algorithms needed to perform successful unsupervised learning.
13. TRUE/FALSE - Humans have a fundamental competence lacking in current AI systems: common sense. We have vast background knowledge of the world, both its physical and social aspects. We have a good sense of how objects - both animate and living - are likely to behave, and we use this knowledge extensively in making decisions about how to act in any given situation.
14. TRUE/FALSE - Many people believe that until AI systems have common sense as humans do, we won't be able to trust them to be fully autonomous in complex real-world situations.
15. TRUE/FALSE - Superficial changes to images, such as slightly blurring or speckling an image, changing some colors, or rotating objects in the scene, can cause ConvNets to make significant errors even when these perturbations don't affect humans' recognition of objects. This unexpected fragility of ConvNets – even those that have been said to “surpass humans at object recognition” – indicates that they are overfitting to their training data and learning something different from what we are trying to teach them, a phenomenon that results in various manifestations of unreliability.
16. The unreliability of ConvNets can result in embarrassing – and potentially damaging – errors. Select a particularly embarrassing/damaging example of unreliability in ConvNets, and de-

scribe it in just a sentence or two.

17. At the end of the section on biased AI, MM observes that the problem of bias in applications of AI has been getting a lot of attention recently, with many articles, workshops, and even academic research institutes devoted to this topic. What questions does she raise in conjunction with this observation? What do you think are the appropriate answers to these questions?
18. TRUE/FALSE - You can often trust that people know what they are doing if they can explain to you *how* they arrived at an answer or a decision. However, “showing their work” is something that deep neural networks – the bedrock of AI systems – cannot easily do.
19. TRUE/FALSE - Recall that a convolutional neural network decides what object is contained in an input image by performing a sequence of mathematical operations (convolutions) propagated through many layers. For a reasonably sized network, these can amount to billions of arithmetic operations. While it would be easy to program the computer to print out a list of all the additions and multiplications performed by a network for a given input, such a list would give us humans *zero* insight into how the network arrived at its answer. A list of a billion operations is not an explanation that a human can understand.
20. What, according to MIT’s *Technology Review* is **the dark secret at the heart of AI**?
21. What does the phrase “theory of mind” refer to, and how is it related to our interactions with AI systems such as deep networks?
22. One of the hottest new areas of AI is variously called “explainable AI,” “transparent AI,” or “interpretable machine learning.” To what do these terms refer?
23. The field of “adversarial learning” has emerged in response to the fact that AI systems can readily be fooled in dramatic fashion, like mixing up a guy in glasses with Milla Jovovich, or misclassifying a stop sign for a speed-limit sign. Briefly describe the field of **adversarial learning**.
24. Jeff Clune, an AI researcher at the University of Wyoming, made a very provocative analogy when he noted that there is “a lot of interest in whether Deep Learning is ‘real intelligence’ or a ‘Clever Hans.’” Explain the essential question that underlies this analogy, being sure to incorporate a few words on the actual Clever Hans.