

# Quiz 1

Started: Jan 18 at 9:33pm

## Quiz Instructions

### Intro and Universal Approximators

This quiz covers lectures 1 and 2. Several of the questions invoke concepts from the hidden slides in the slide deck, which were not covered in class. So please go over the slides before answering the questions.

You will have three attempts for the quiz. Questions will be shuffled and you will not be informed of the correct answers until after the deadline. While you may discuss the concepts underlying the questions with others, you must solve all questions on your own - see course policy.



Question 1 1 pts

Which of your quiz scores will be dropped?

Hint: watch Lecture 0



Lowest 3 quiz scores



No scores will be dropped



Lowest 2 quiz scores



Lowest 1 quiz scores



Question 2 1 pts

Is the following statement true or false? Hebbian learning allows reduction in weights and learning is bounded.

Slide: lec 1, "Hebbian Learning" slide 66



True



False



### Question 3 1 pts

Match the corresponding terms and definitions introduced in Lecture 1.

Hint: Lecture 1: Slides on 31-81

The McCulloch and Pitts model

[ Choose ] ▾

Alexander Bain

[ Choose ] ▾

Lawrence Kubie

[ Choose ] ▾

Hebbian Learning

[ Choose ] ▾

Marvin Minsky and Seymour Papert

[ Choose ] ▾

One of David Hartley's Observations

[ Choose ] ▾

Frank Rosenblatt

[ Choose ] ▾

Associationism Theory by Aristotle

[ Choose ] ▾



### Question 4 1 pts

How does the number of weights (note: not neurons) in an XOR network with **threshold logic** perceptrons with **1 hidden layer** grow with the number of inputs to the network?

Hint: Review Lec 2: Slides on “Caveat 2” (Slide 75)



Between polynomial and exponential



Linear



Exponential or faster



Polynomial but faster than linear



Question 5 1 pts

How does the number of weights (note: not neurons) in an XOR network with *threshold logic* perceptrons with *1 hidden layer* grow with the number of inputs to the network?

See *lec 2: Slides on “Optimal depth” and “Network size” 113-123*



Linear



Between polynomial and exponential



Exponential or faster



Polynomial but faster than linear



Question 6 1 pts

Suppose the data used in a classification task is 10-dimensional and positive in all dimensions. You have two neural networks. The first uses threshold activation functions for hidden layers, and the second uses softplus activation functions for hidden layers. In both networks, there are 2000 neurons in the first hidden layer, 8 neurons in the second hidden layer, and a huge number of neurons for all later layers. It turns out that the first network can never achieve perfect classification. What about the second network?

Hint: A layer with 8 neurons effectively projects the data onto a 8-dimensional surface of the space. The input is 10-dimensional.

Lec 2 slides 138-150



It might fail for some data sets, since the 8 neurons in the second hidden layer could bottleneck the flow of information. In that case, the sizes of layers 3 and above don't matter.



Assuming that layers 3 and above are so expressive that they never bottleneck the flow of information, the second network will be able to achieve perfect classification.



It might fail for some data sets, not only because a bottleneck can occur, but also because the 2000 neurons in the first hidden layer could bottleneck the flow of information if the classification task is sufficiently complex.

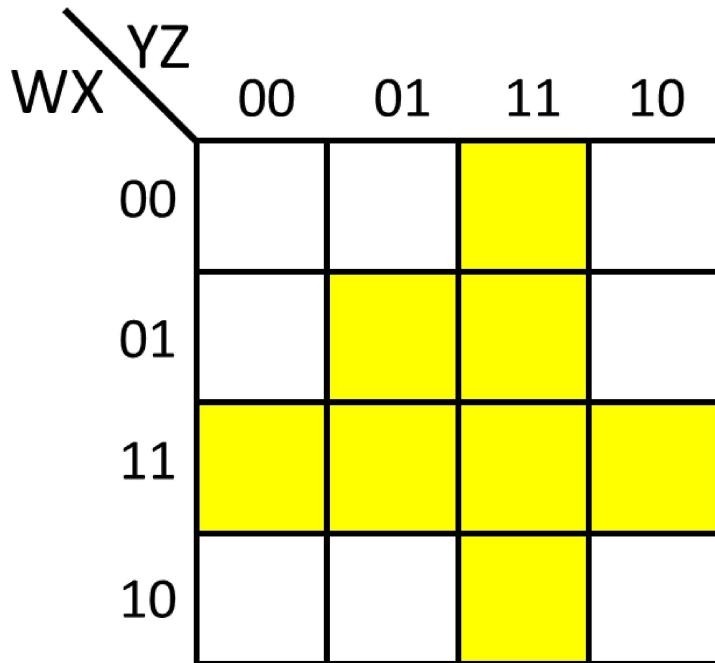


It too is guaranteed to be unable to achieve perfect classification.



## Question 7 1 pts

**What is the fewest number of neurons needed (including any output layer neurons) for a network to implement the truth table shown by the following Karnaugh map? (numeric answer, int and float are both fine)**

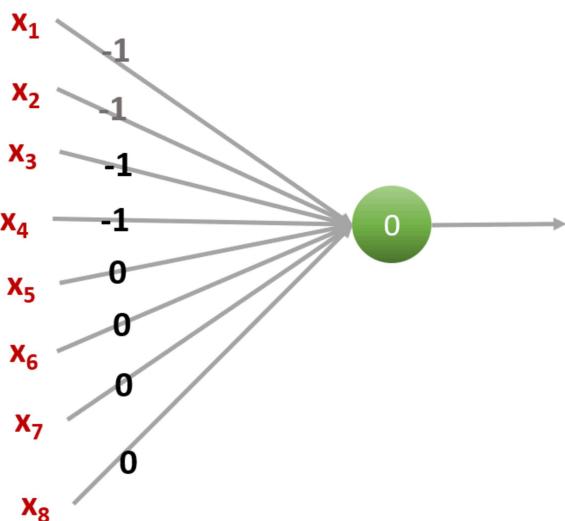


*Hint: lec 2, “Reducing a Boolean Function”. Slide 47 - 50*



## Question 8 1 pts

Under which conditions will the perceptron graph below fire? Note that  $\sim$  is NOT. (select all that apply)



## Slide: lec 2, "Perceptron as a Boolean gate", slides 26-30



fires only if  $x_1, x_2, x_3, x_4$  are all 0, regardless of  $x_5 \dots x_8$



$x_1 \& x_2 \& x_3 \& x_4$



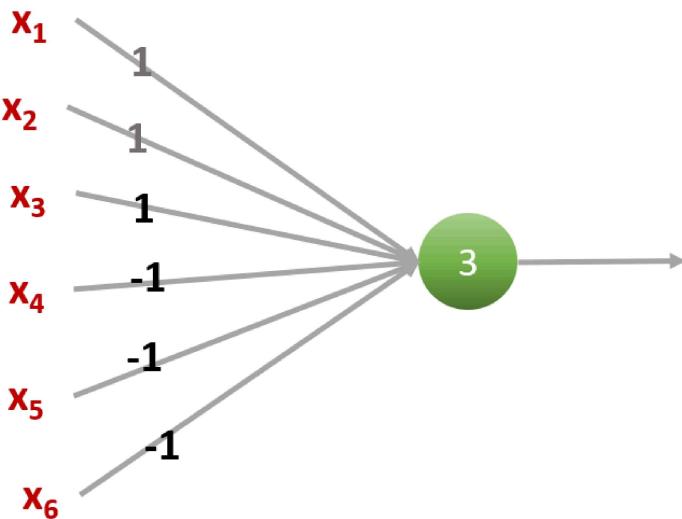
$\sim x_1 \& \sim x_2 \& \sim x_3 \& \sim x_4$



Never fires



Question 9 1 pts



Under which condition(s) is the perceptron graph above guaranteed to fire? Note that  $\sim$  is NOT. (select all that apply)

## Slide: lec 2, "Perceptron as a Boolean gate" slides 26-30



$x_1 \& x_2 \& x_3 \& \sim x_4 \& \sim x_5 \& \sim x_6$



$\sim x_1 \& \sim x_2 \& \sim x_3 \& x_4 \& x_5 \& x_6$



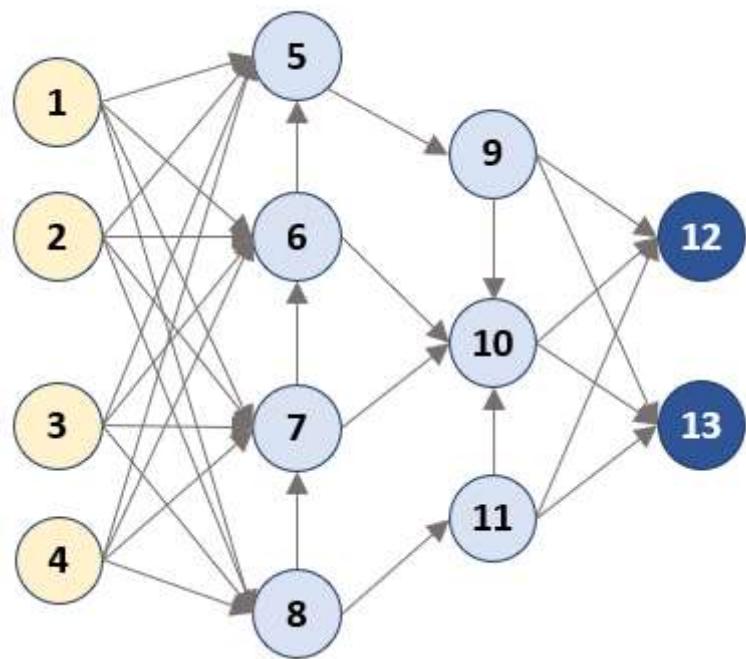
Never fires



$x_1 \& \sim x_2 \& x_3 \& \sim x_4 \& x_5 \& \sim x_6$



Question 10 1 pts



If the yellow nodes are inputs (not neurons) and the dark blue nodes are outputs, which neurons are in layer 6?

Hint: lec 2, slides 19-20 on "What is a layer"

- 10, 12
- 10
- 9, 10, 11
- 10, 12, 13

Quiz saved at 9:33pm

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