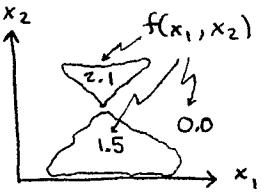


5 Apr 1990
Neil E Cotter

Approximation Theory - Universal Approximation - Perceptron in Computing arbitrary functions

ref: Lippmann article handed out in class

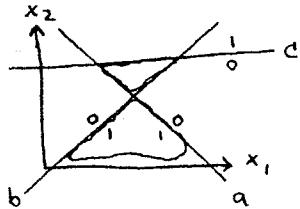
Suppose we wish to create a perceptron network to compute the following function (plotted as a topographic map):



Function $f(x_1, x_2)$ has two triangular mesas of height 1.5 and 2.1.

We use a three-layer network to approximate $f()$:

- Layer 1) Use perceptrons (3 in this case) to outline polygons that approximate the boundaries of the mesas.



For greater accuracy, use more lines.

- Layer 2) Use perceptrons (2 in this case) as logic gates (usually AND gates) to determine if input (x_1, x_2) to net lies in a given polygon, (i.e. is on the correct side of all lines forming polygon). Thus, we have 1 logic gate for each polygon. Gate output is 1 if and only if point (x_1, x_2) is inside that gate's polygon. Only one gate output will be 1 at a given time.

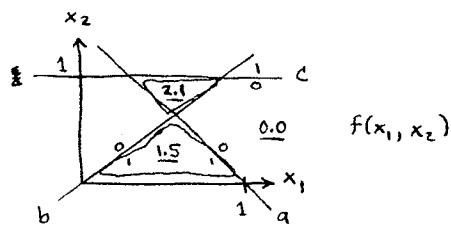
- Layer 3) Use a perceptron from which the step function has been removed. This neuron is called a summing node or linear neuron. Its inputs are logic gate outputs which are 0 or 1. Set the synapse (receiving a given logic gate output) to the value or height of $f()$ in the corresponding polygon, here (1.5 or 2.1).

Approximation Theory - Universal Approximation -
 Perceptron \approx Computing arbitrary functions (cont.)

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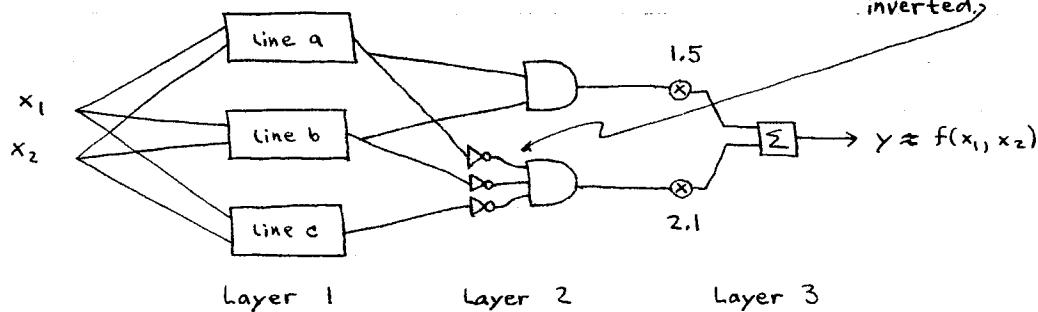
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ex:



$$f(x_1, x_2)$$

Note that 2.1 region
is on "0" side of
all three lines. This
is why AND gate inputs
inverted,



Detailed view:

