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## 2.1

### a

A nonlinear transfer function gives the neural network a universal property: Given enough layers and neurons, the network can model any function within a certain accuracy. In a network with a linear transfer function we can only compute a linear function. A network with a linear transfer function of  $n$  layers will always be equivalent to a network with only one layer: linear functions can always be concatenated.

Whenever the function that we are trying to model isn't a linear function, it is useful to use a nonlinear transfer function. Examples include image classification or speech recognition.

### b

Consider a simple neural network with two input neurons that can both either be 0 or 1 and one output layer. We want to construct an AND gate with our network, so without bias our quest would be to find  $w = (w_1, w_2)$  such that:

$$\begin{aligned}0 &\leq 0 \\ w_1 &\leq 0 \\ w_2 &\leq 0 \\ w_1 + w_2 &> 0\end{aligned}$$

which is impossible. We can easily however create the network with a bias, if we have the weights  $w = (1, 1)$  and the bias  $\theta = \frac{3}{2}$ . Then  $\text{sgn}(w^T x - \theta)$  would give us AND.

### c

Point and edge filters are for example a connectionist neuron which gets values of a scalar field as input, that represent the color of each pixel or the color gradient or even a higher derivative and has weights in the following form: In the simple case of two colors (0 and 1) this point filter would return zero for no point, 1 or -1 for a point in the outer region and 8 or -8 if the point is in the middle. This goes analogously for the other filter.

-1	-1	-1
-1	+8	-1
-1	-1	-1

point filter

-1	0	+1
-2	0	+2
-1	0	+1

edge filter (Sobel filter)

+1	+2	+1
0	0	0
-1	-2	-1

**d**

The first is deterministic and the second has a noise parameter and can return different states for set parameters and a given input.

## 2.2

**a**

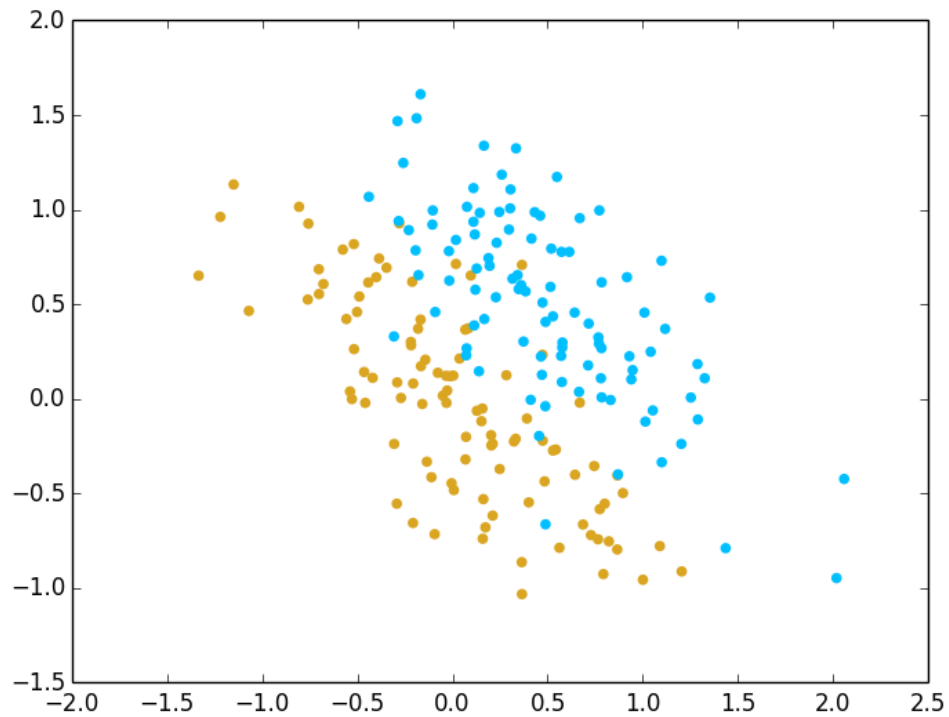


Figure 1: Plot where Y=1 (blue) and Y=0 (gold).

**b**

**c**

Best weight vector is .

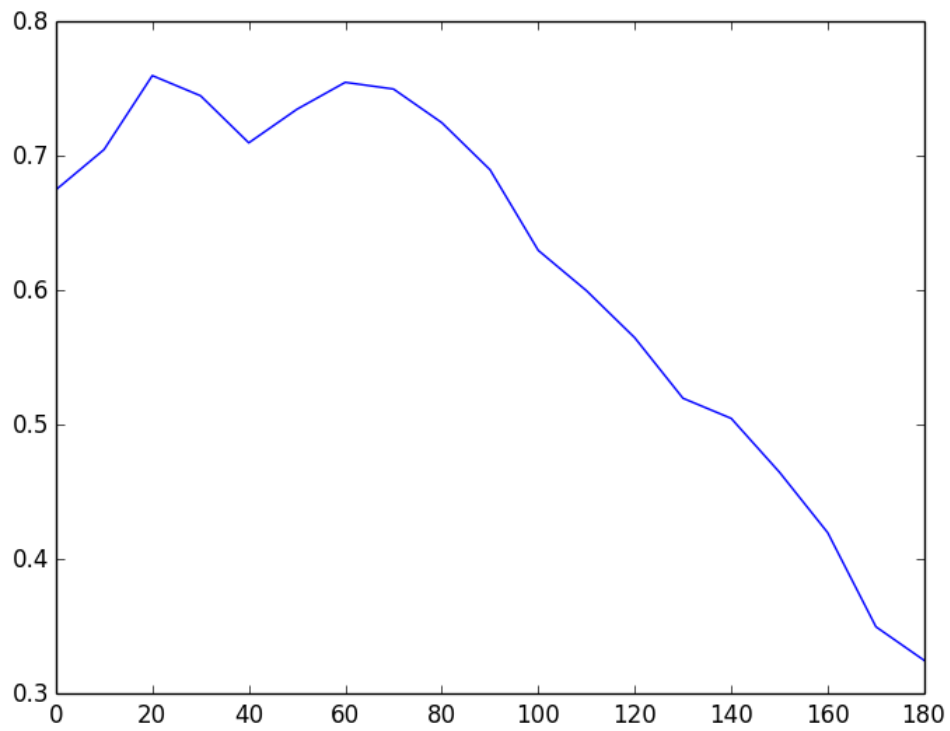


Figure 2: Classification performance for different alphas.

**d**

**e**

## **2.3**

**a**

A MLP could decide between a horizontal and a vertical edge, whereas a perceptionist neuron would either be able to differ between vertical edge or no vertical edge OR horizontal edge or no horizontal edge.

**b**

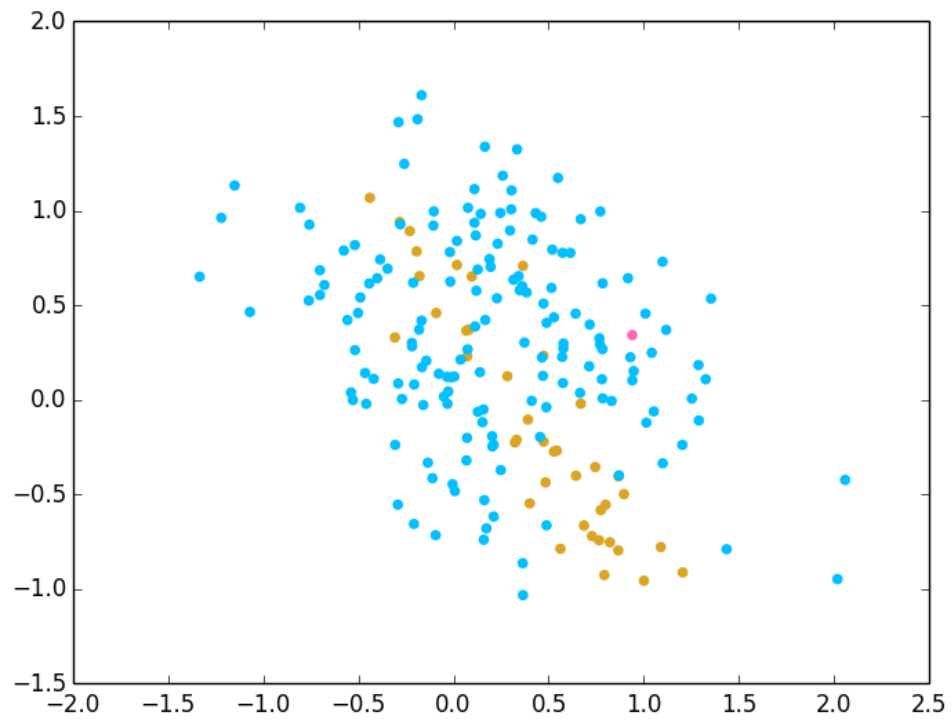


Figure 3: Classified data points. Blue = correctly classified, Gold = other data points, Pink = weight.

**c + bonus**

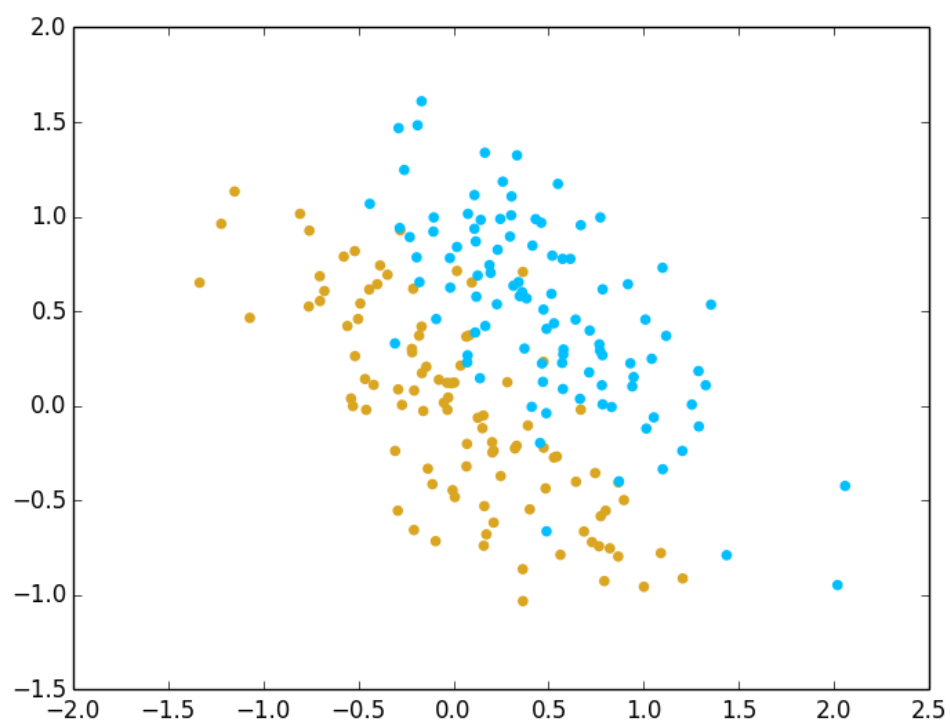


Figure 4: Plot where  $Y=1$  (blue) and  $Y=0$  (gold).

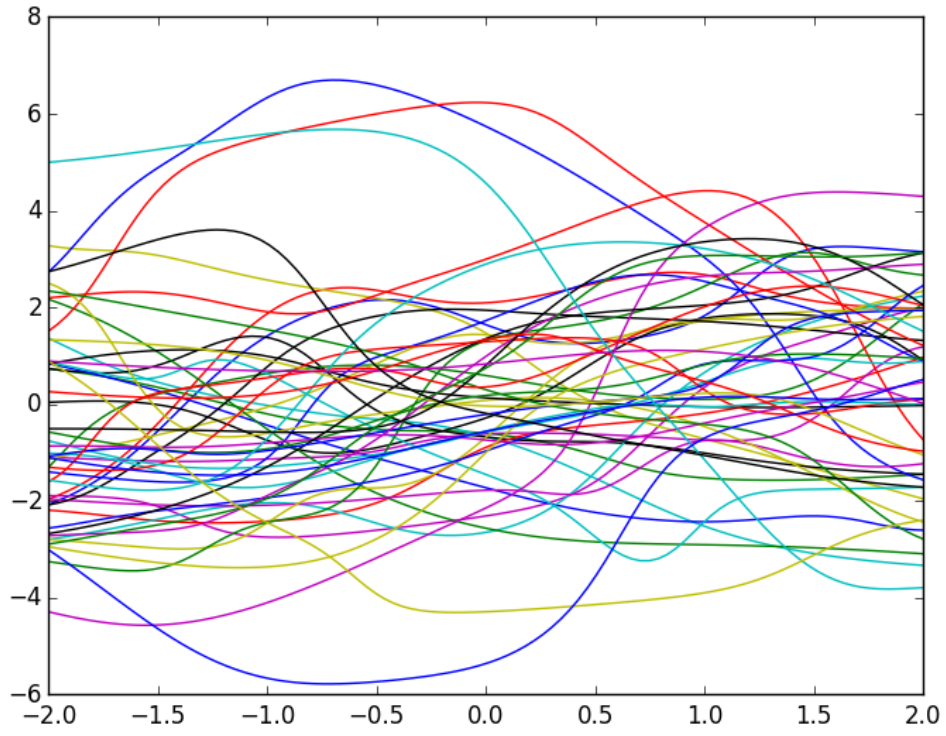


Figure 5: functions computed with normally distributed  $a_i$  with a standard deviation of 2

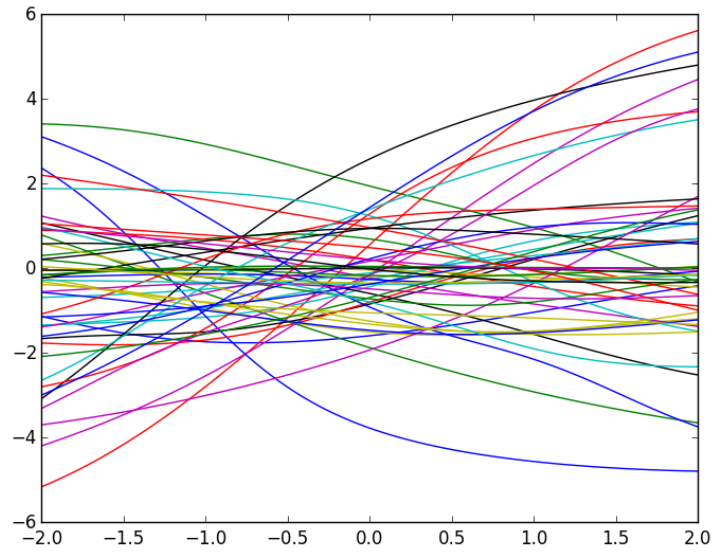


Figure 6: functions computed with normally distributed  $a_i$  with a standard deviation of 0.5

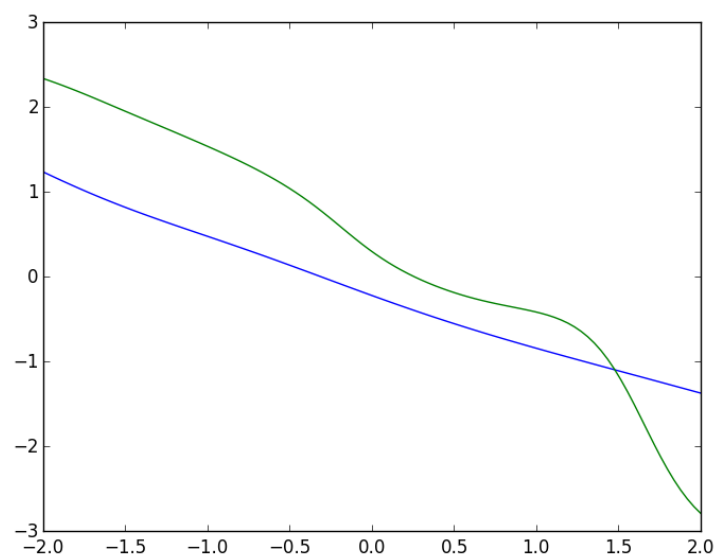


Figure 7: functions with least mean square error from  $f(x)=-x$ , computed with an std of 2 (green) and 0.5 (blue)