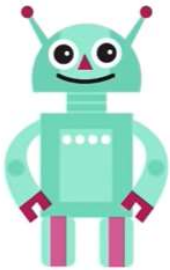
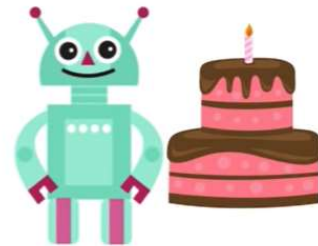


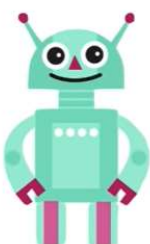
INTRODUCTION TO DEEP LEARNING:



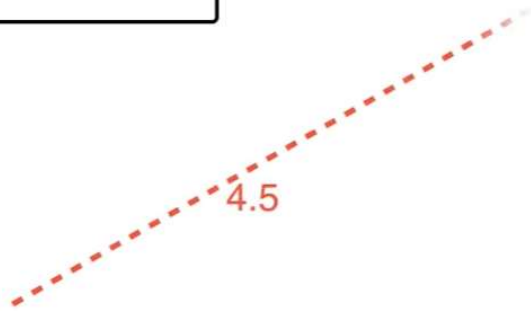
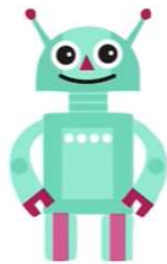
- turn right
- go 10 steps
- turn left
- go 4 steps
- grab cake



- calculate distance to cake
- move around, pick a direction where the distance decreases



- calculate distance to cake
- move around, pick a direction where the distance decreases
- take a step in that direction



- calculate distance to cake
- move around, pick a direction where the distance decreases
- take a step in that direction
- repeat



Gradient descent

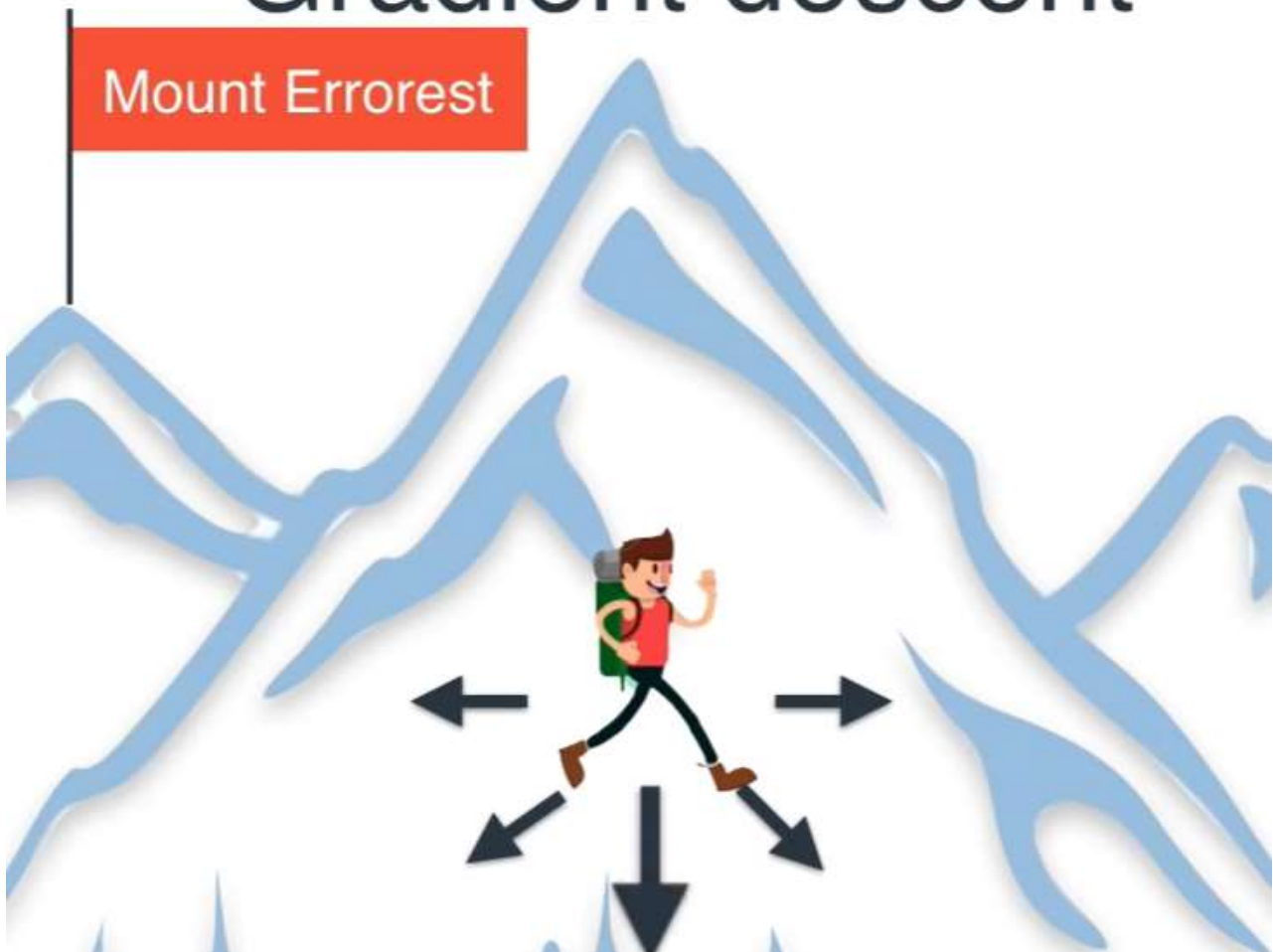
Mount Errorrest



Gradient descent



Gradient descent



WE TOOK THE GRADIENT – THE DERIVATIVE

Gradient descent



Get cake
Minimize distance to cake



Descend from mountain
Minimize height



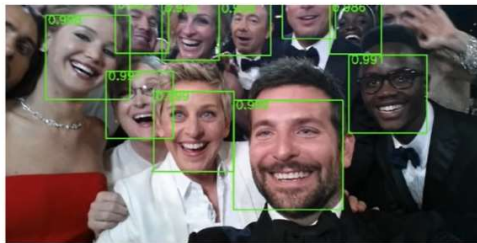
Solve any problem
Minimize error



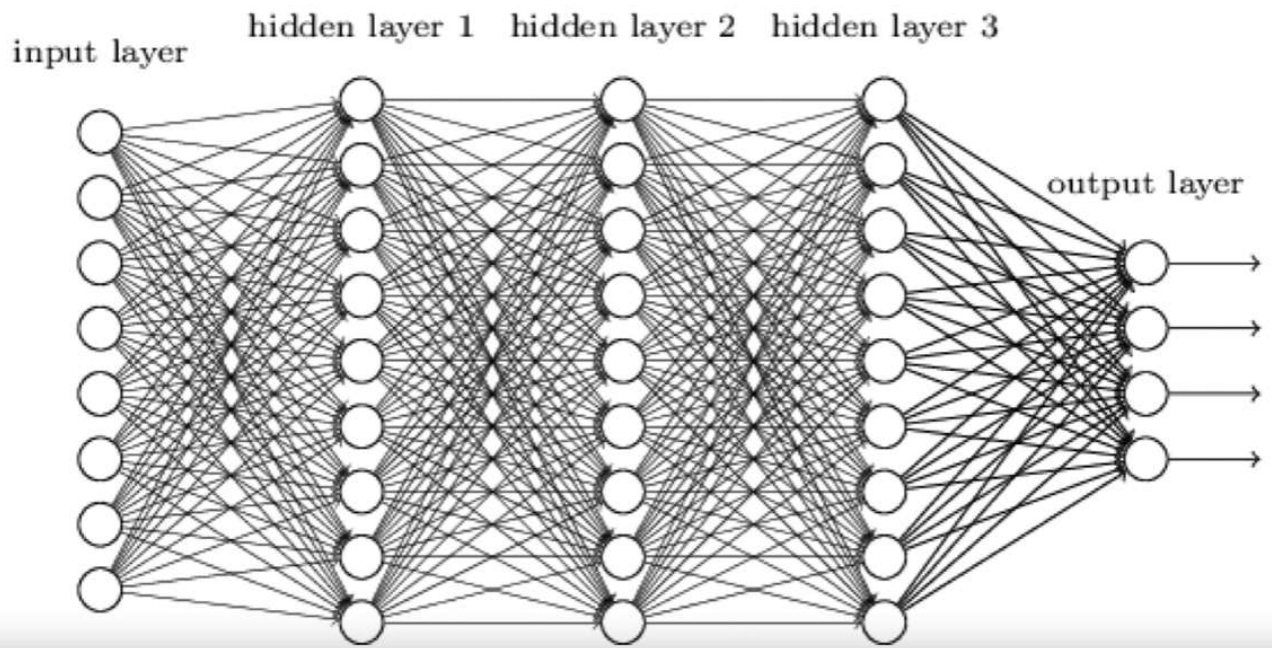
Self Driving Car



Many more things



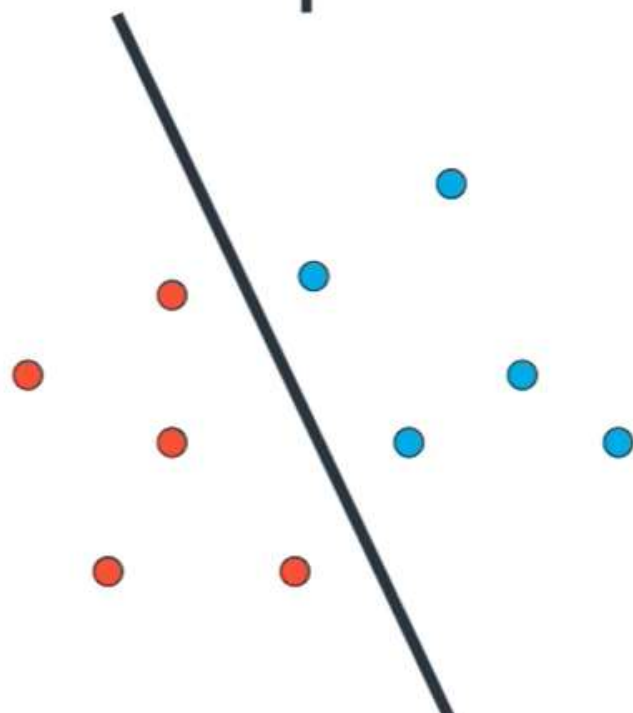
Neural Networks



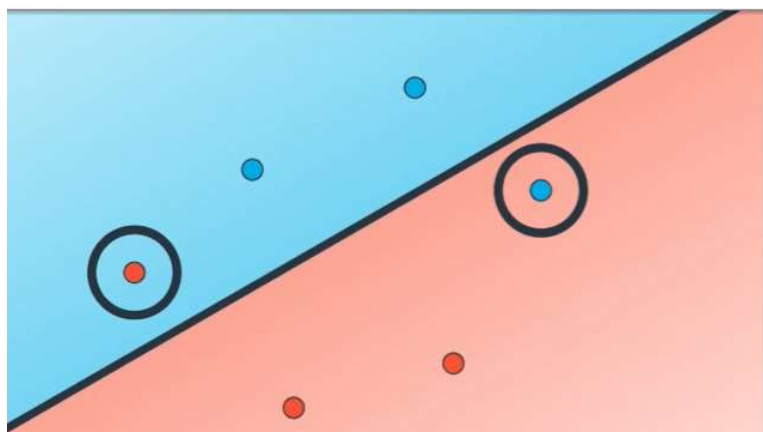
Neural Networks



Goal: Split Data

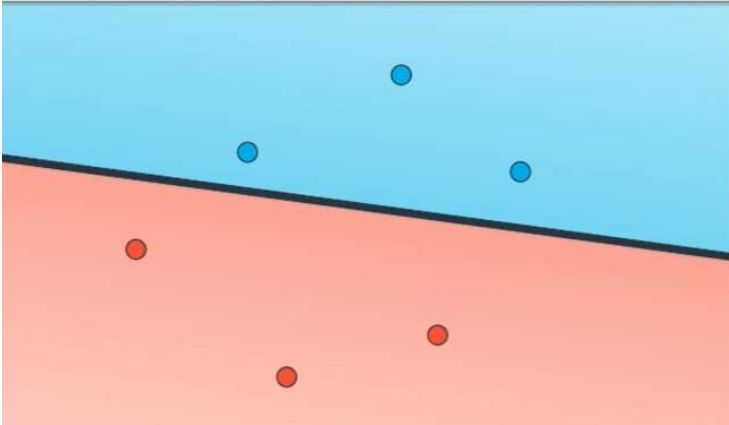


Goal: Split Data



2 errors

Goal: Split Data



0 errors

Hot!

Gradient descent



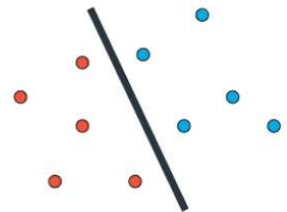
Get cake

Distance to cake
continuous function



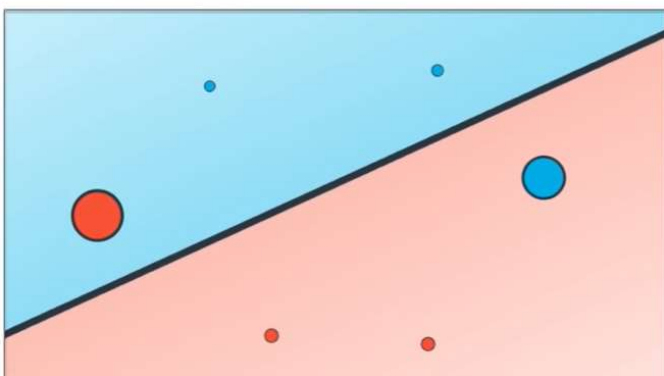
Descend from mountain

Height
continuous function

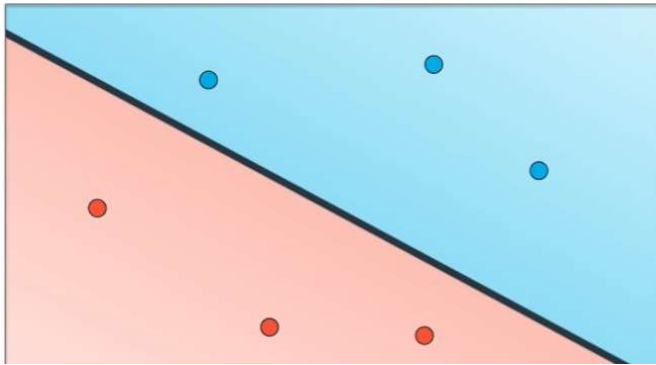


Split data

Number of errors
discrete function



Error = $\bullet + \bullet + \bullet + \bullet + \bullet + \bullet + \bullet$



$$\text{Error} = \text{small blue dot} + \text{small blue dot} + \text{large blue dot} + \text{large red dot} + \text{small red dot} + \text{small red dot} + \text{small red dot}$$

$$\text{Error} = \text{blue dot} + \text{blue dot} + \text{blue dot} + \text{red dot} + \text{red dot} + \text{red dot} + \text{red dot}$$

$$\text{Error} = \text{small blue dot} + \text{small blue dot} + \text{large blue dot} + \text{large red dot} + \text{small red dot} + \text{small red dot} + \text{small red dot}$$

$$\text{Error} = \text{blue dot} + \text{blue dot} + \text{blue dot} + \text{red dot} + \text{red dot} + \text{red dot} + \text{red dot}$$

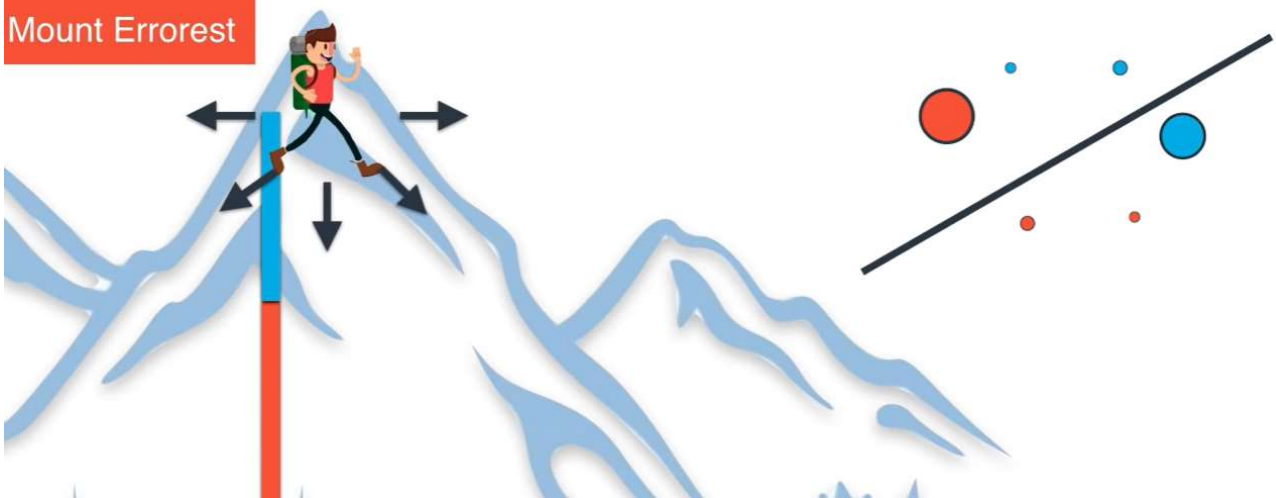
Minimize error



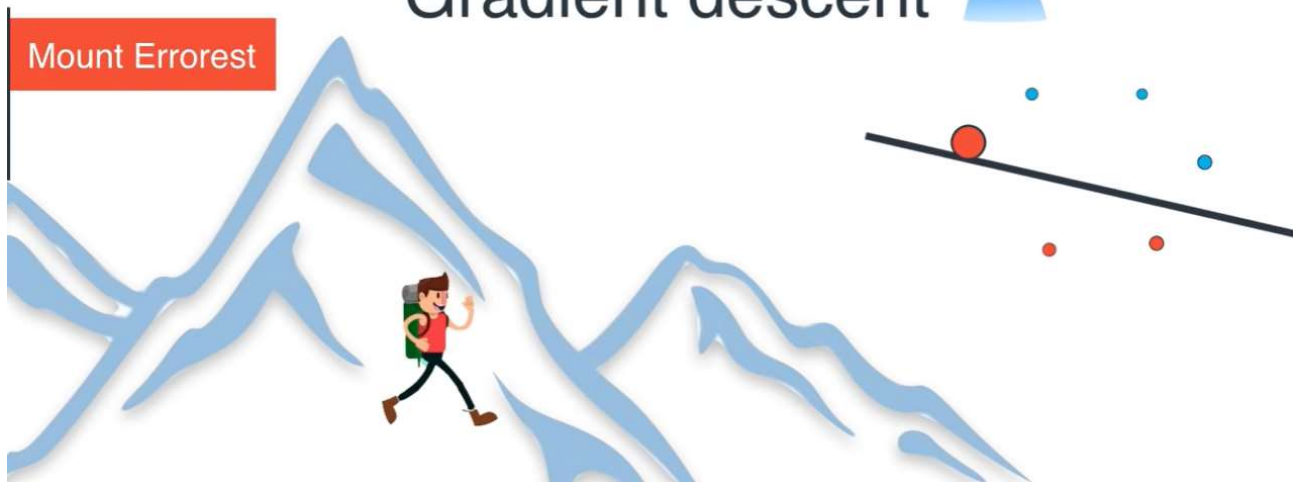
Gradient descent

Gradient descent 

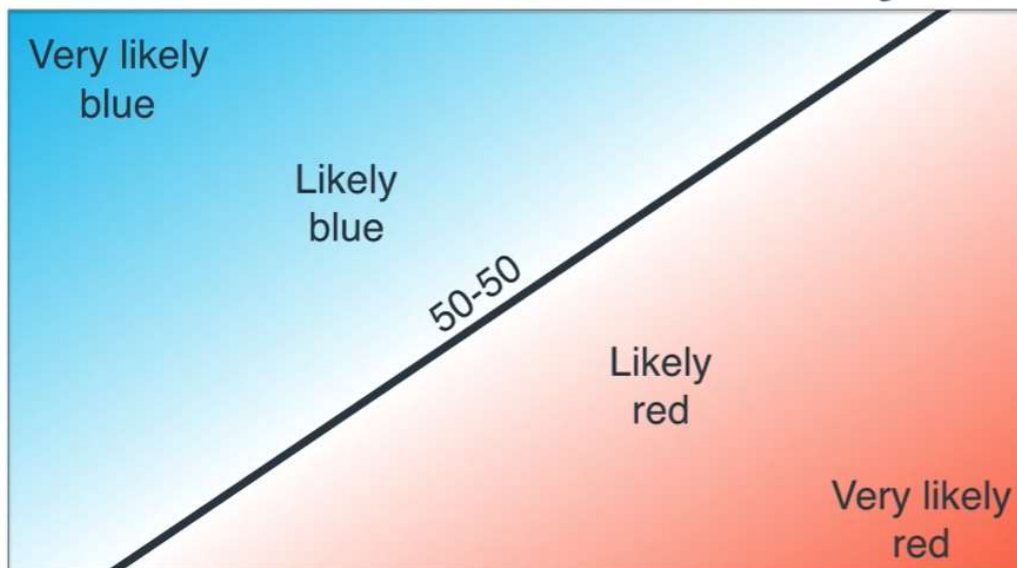
Mount Errorest



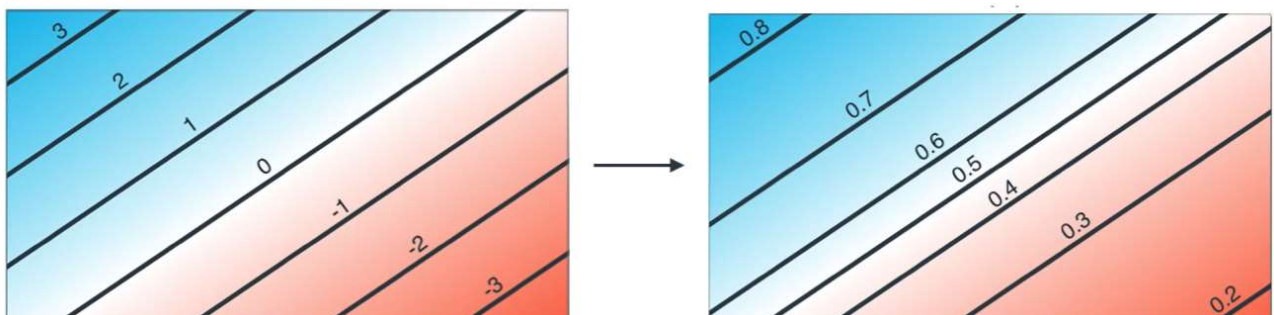
Gradient descent



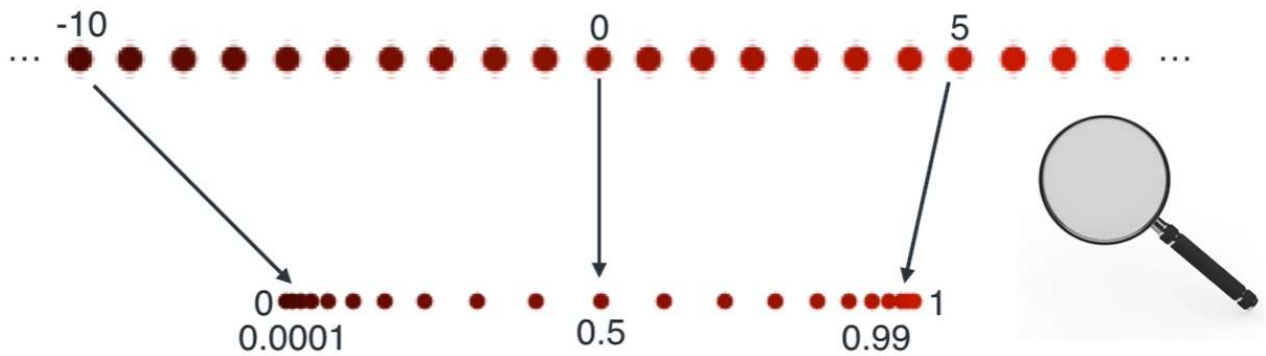
Probability



Probability

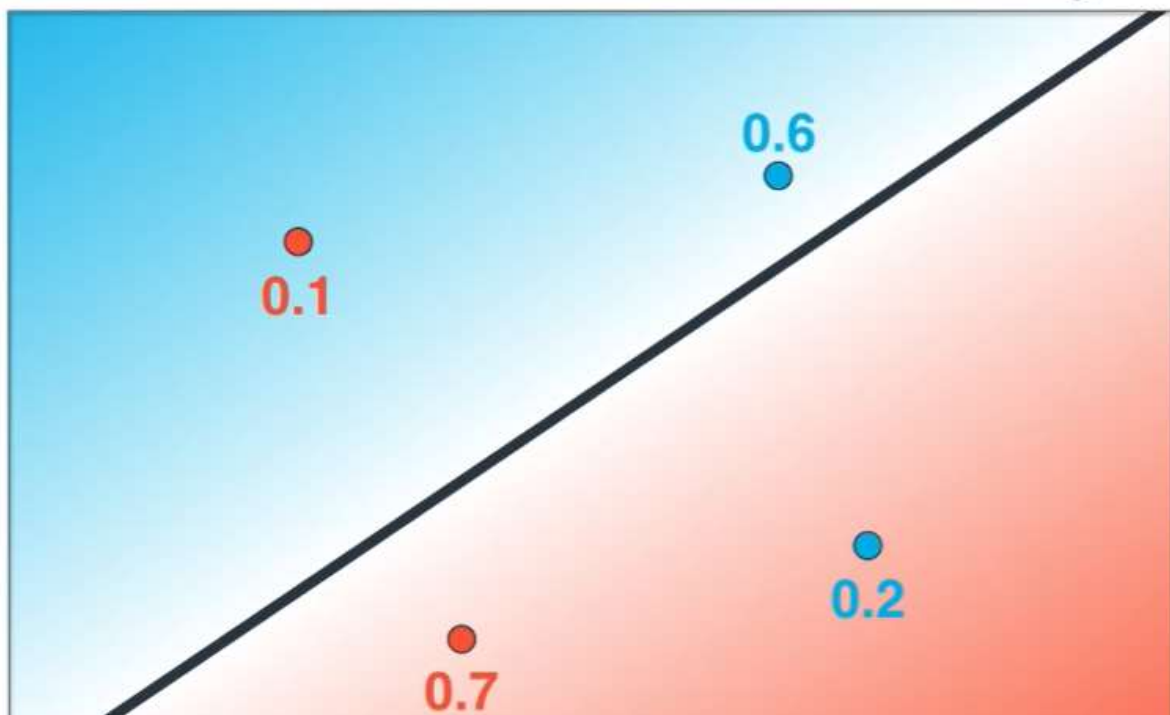


Activation function

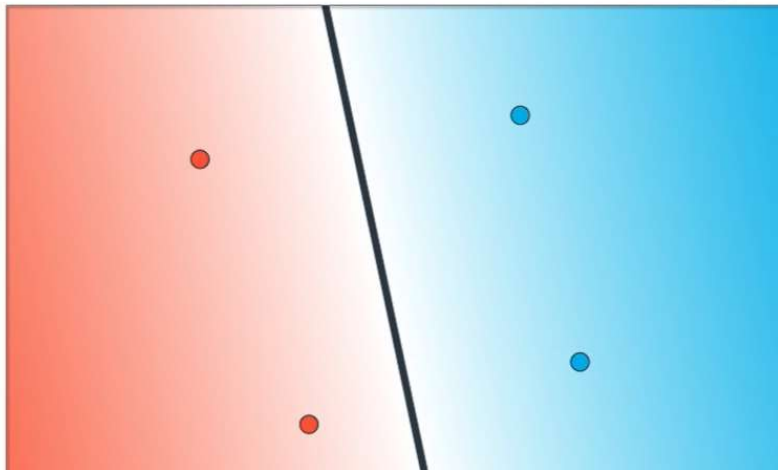


IF WE THINK THESE ARE INDEPENDENT EVENTS, THE PROBABILITY ALL FOUR HAPPENING, IS THE PRODUCT

Probability



Probability

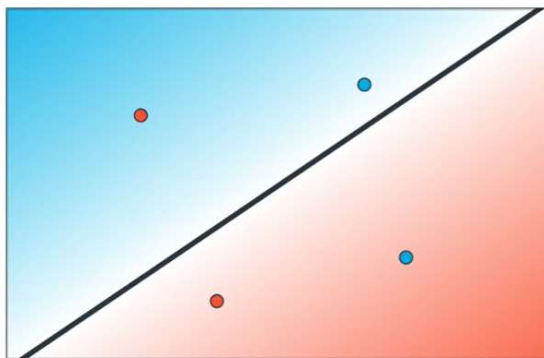


$$0.6 * 0.2 * 0.1 * 0.7 = 0.0084$$

$$0.7 * 0.9 * 0.8 * 0.6 = 0.3024$$

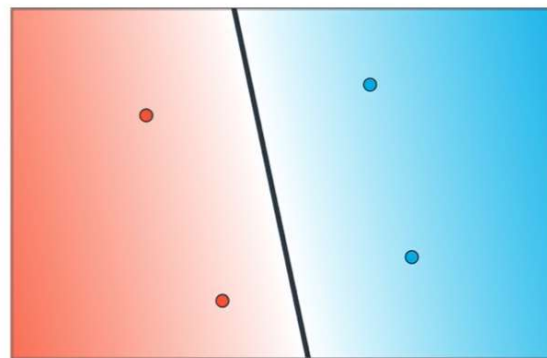
Maximum Likelihood

Error function



$$0.6 * 0.2 * 0.1 * 0.7 = 0.0084$$

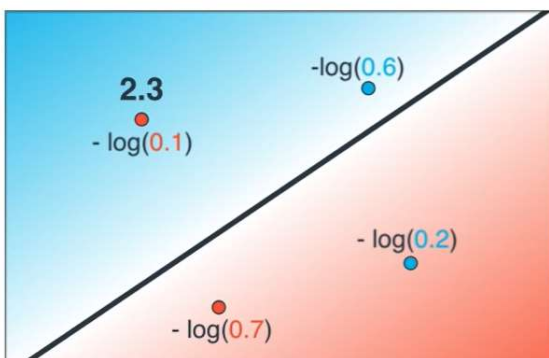
$$-\log(0.6) - \log(0.2) - \log(0.1) - \log(0.7) = 4.8$$



$$0.7 * 0.9 * 0.8 * 0.6 = 0.3024$$

$$-\log(0.7) - \log(0.9) - \log(0.8) - \log(0.6) = 1.2$$

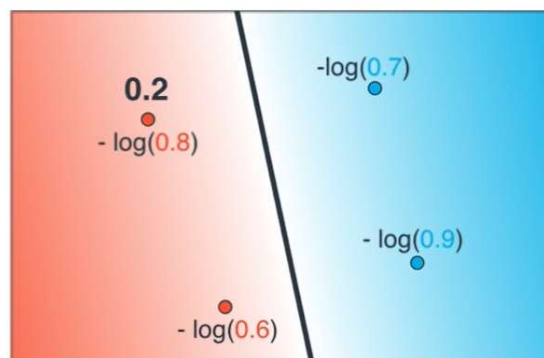
Error function



$$0.6 * 0.2 * 0.1 * 0.7 = 0.0084$$

$$-\log(0.6) - \log(0.2) - \log(0.1) - \log(0.7) = 4.8$$

2.3



$$0.7 * 0.9 * 0.8 * 0.6 = 0.3024$$

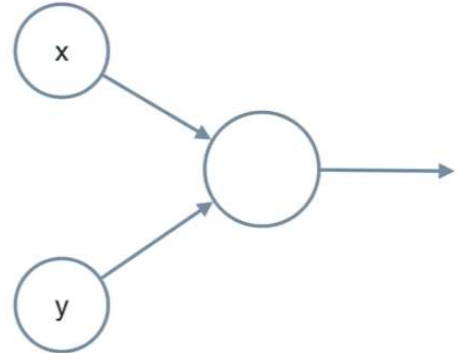
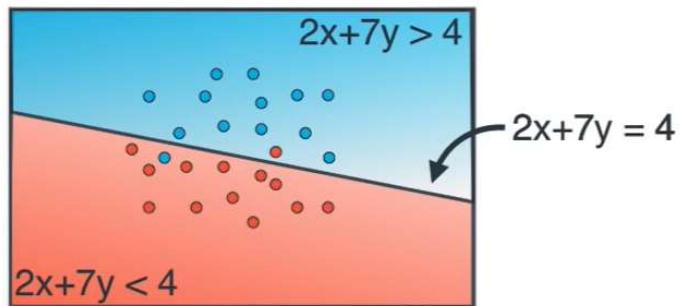
$$-\log(0.7) - \log(0.9) - \log(0.8) - \log(0.6) = 1.2$$

0.2

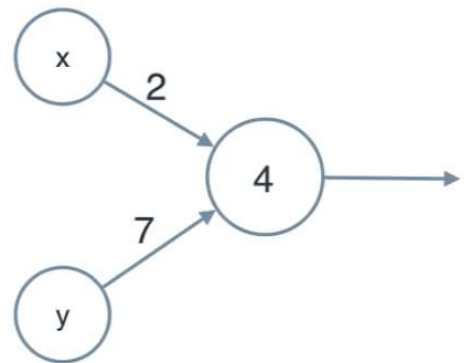
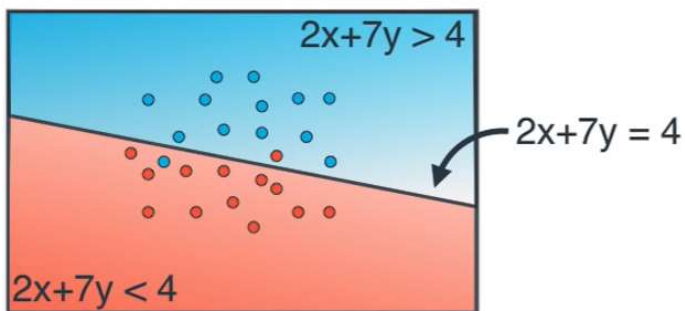


ERROR FUNCTION AS A PENALTY FOR EVERY POINT

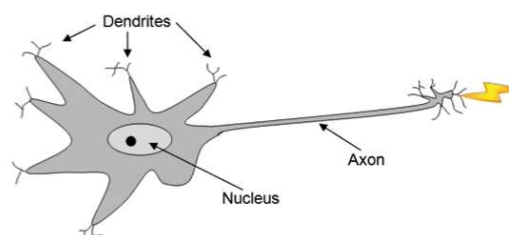
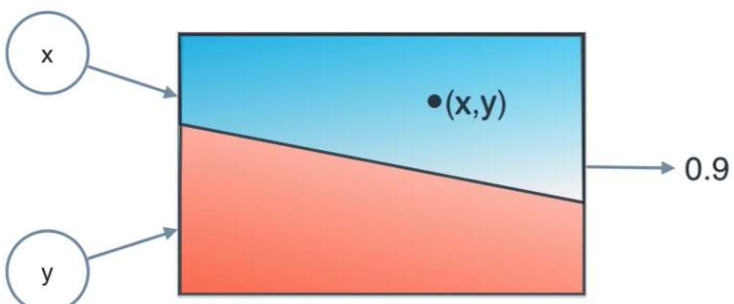
Neuron



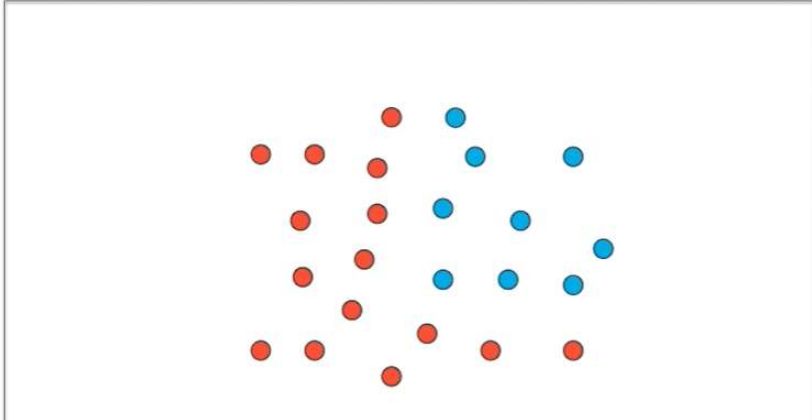
Neuron



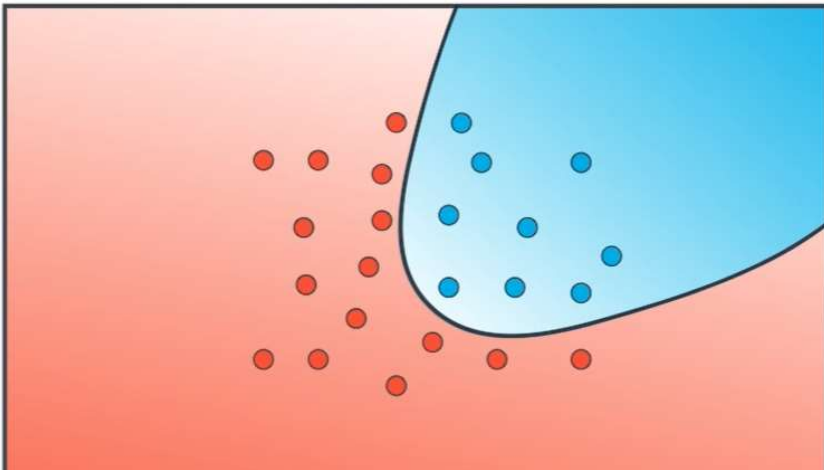
Neuron



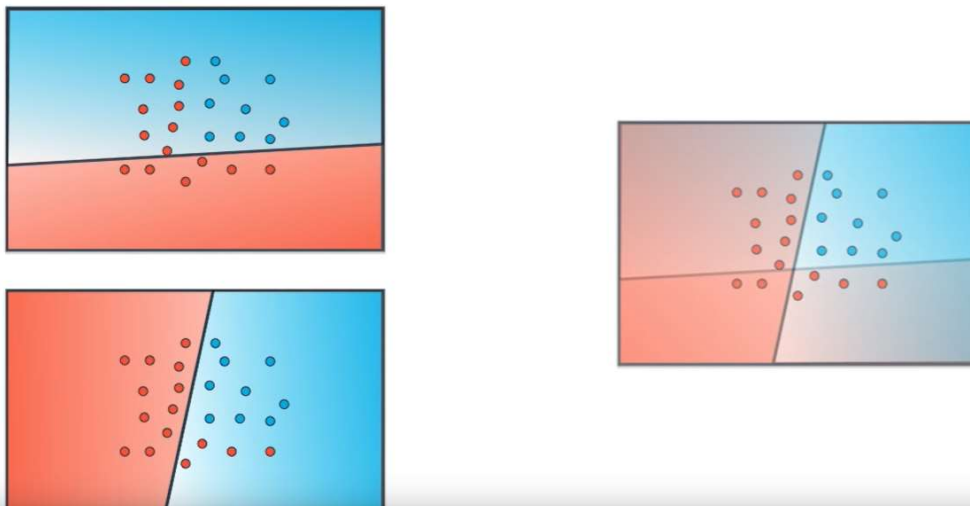
Non-linear regions

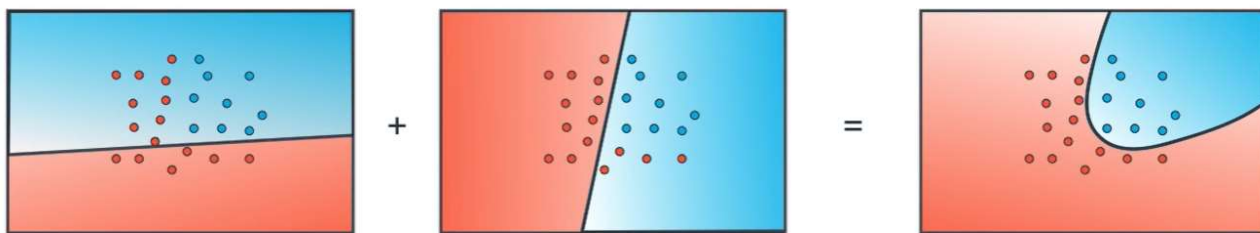


Non-linear regions

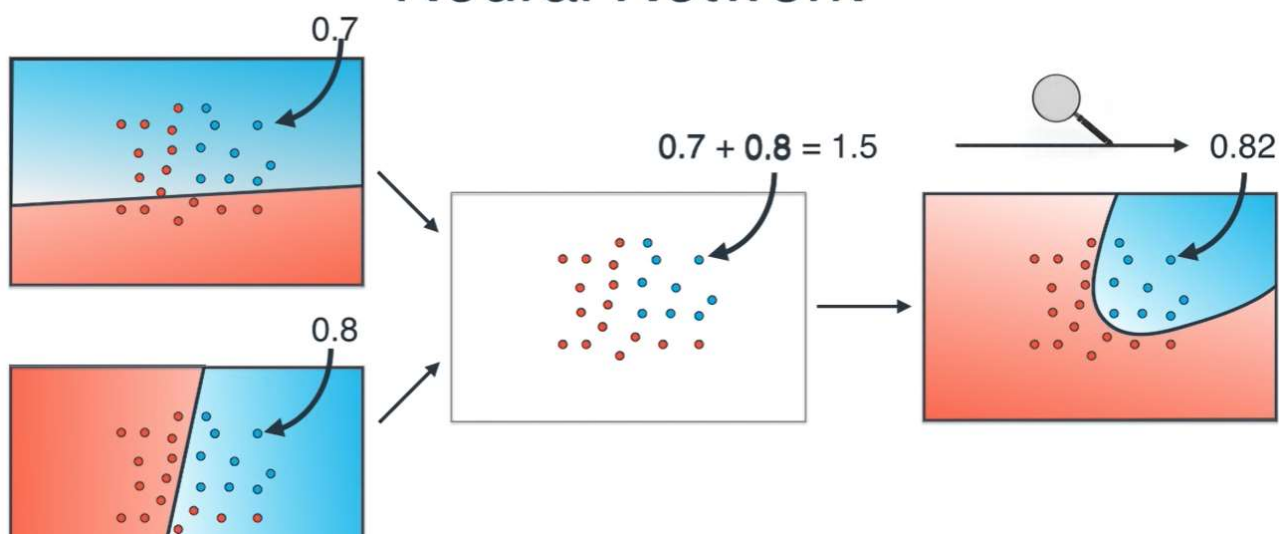


Combining Regions

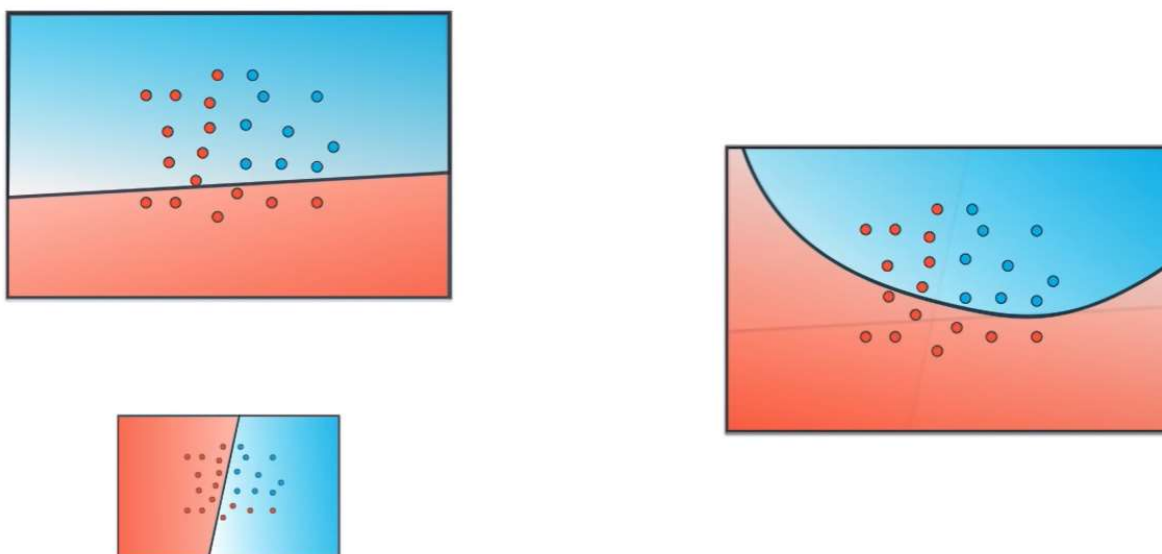




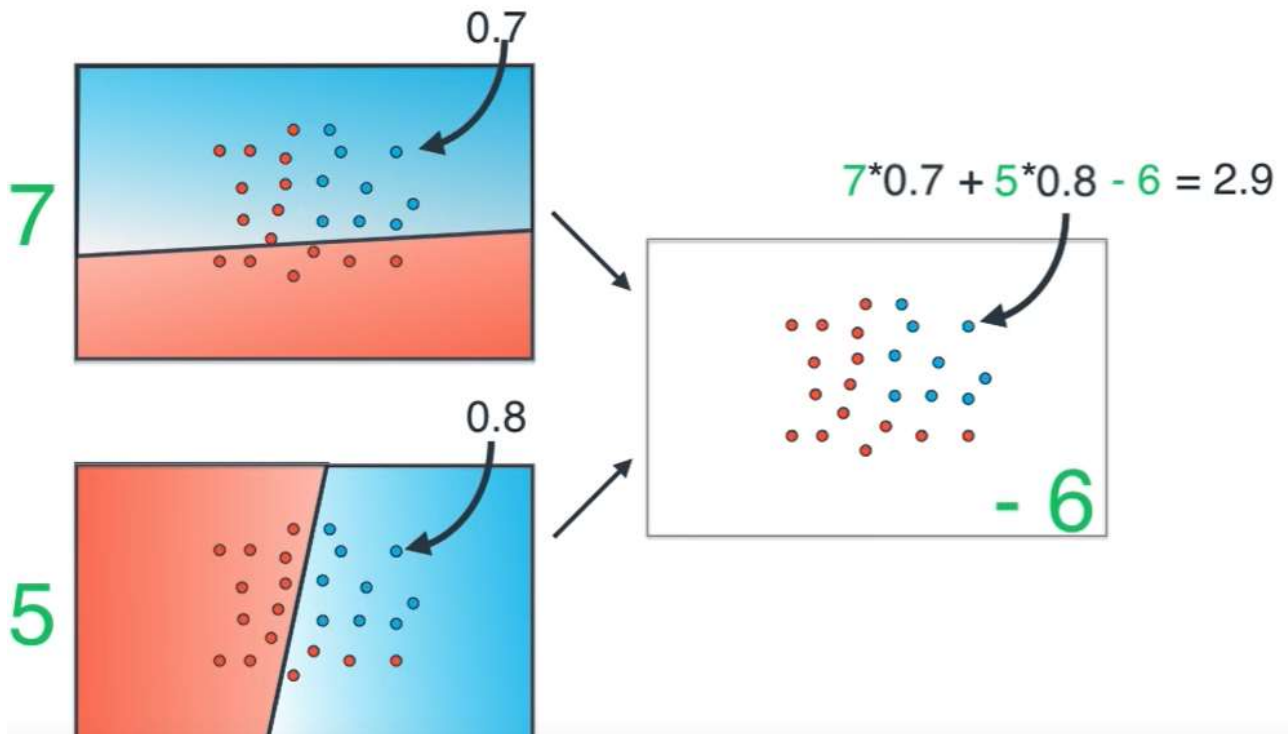
Neural Network



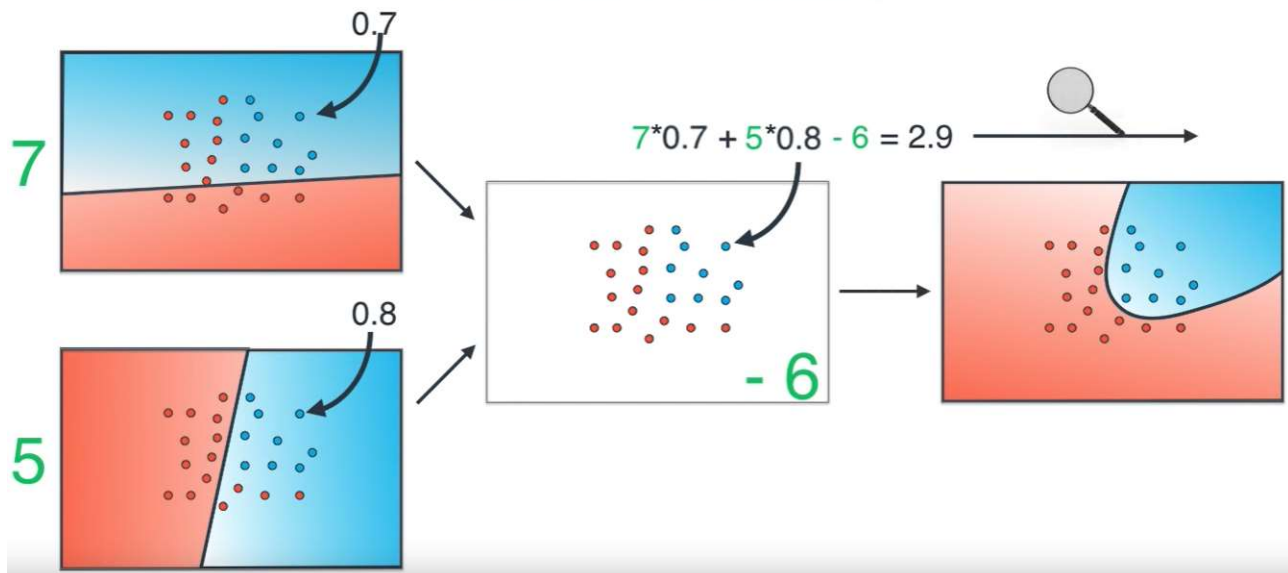
Combining Regions



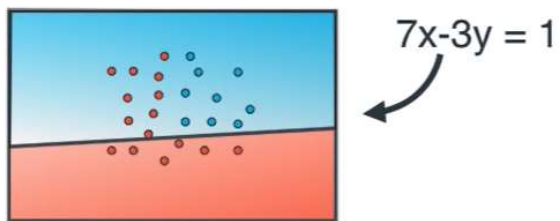
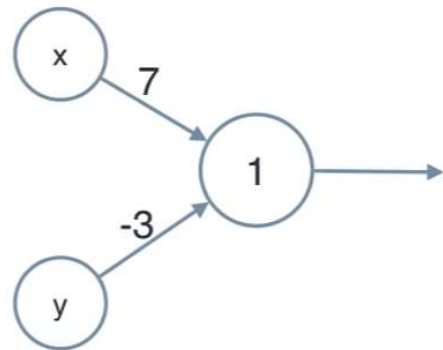
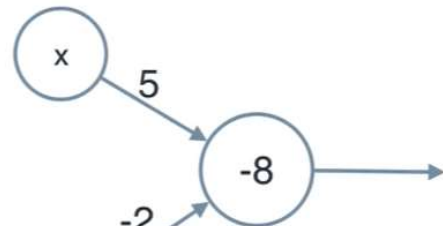
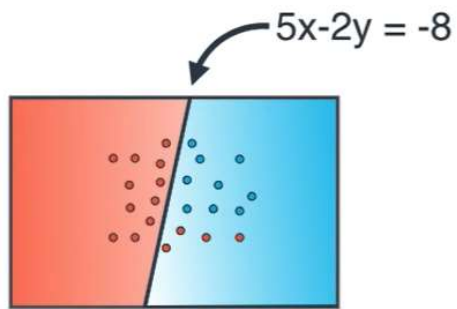
Neural Network



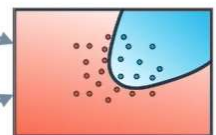
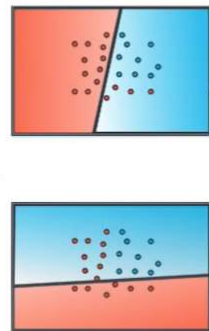
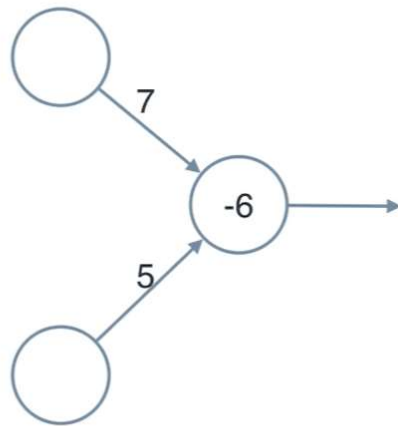
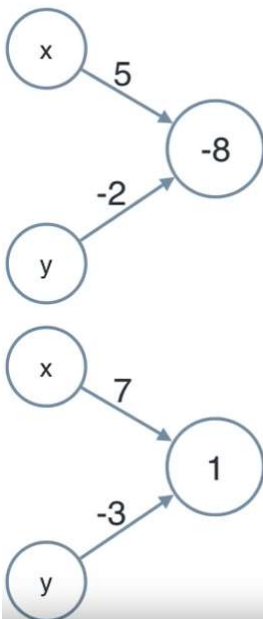
Neural Network



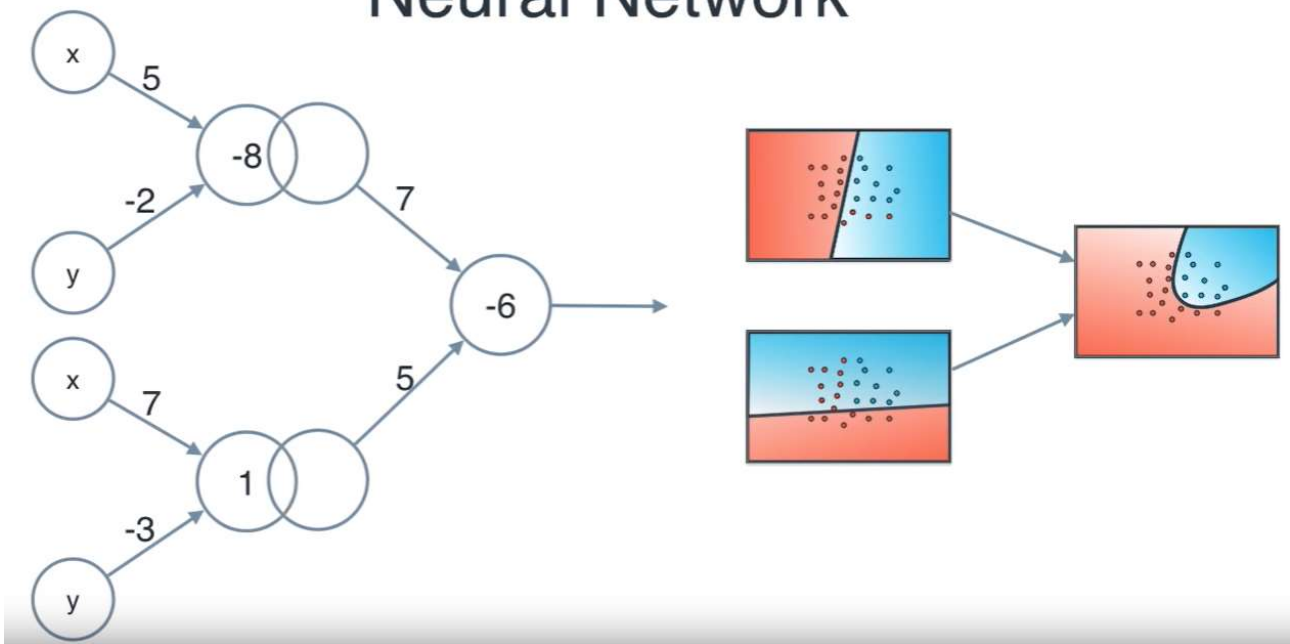
Neural Network



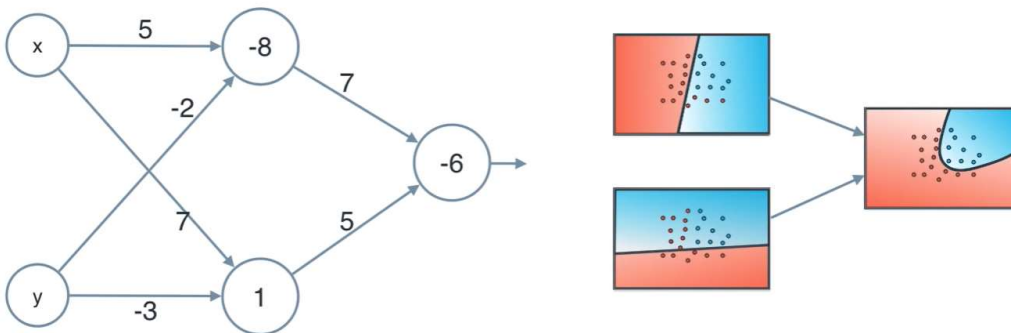
Neural Network



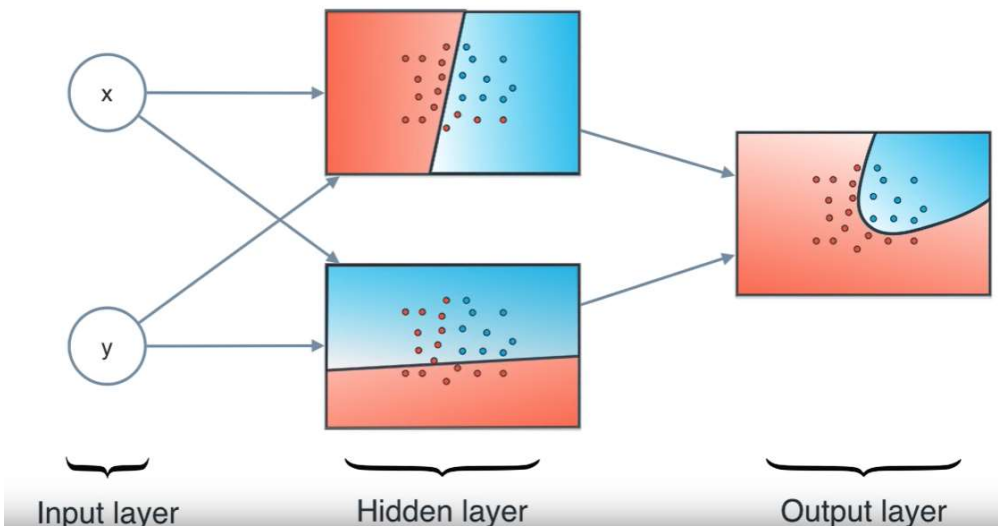
Neural Network

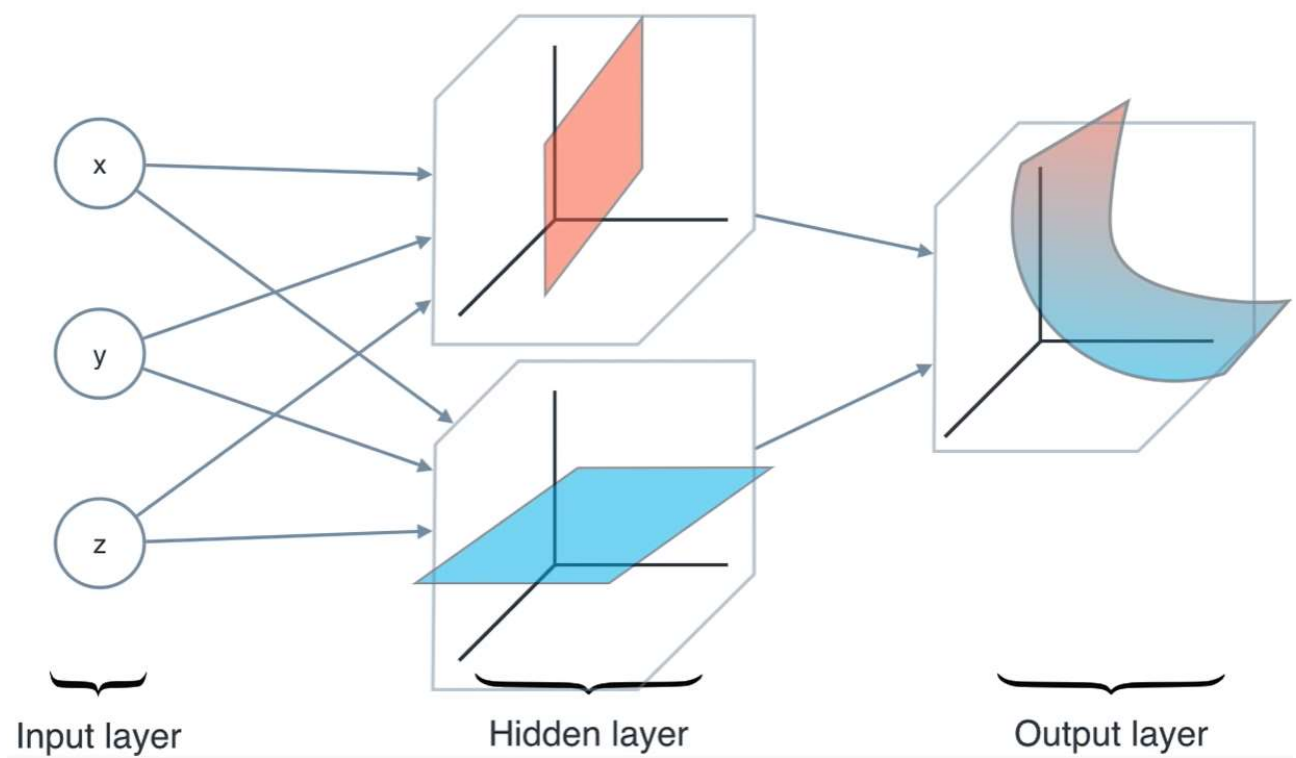
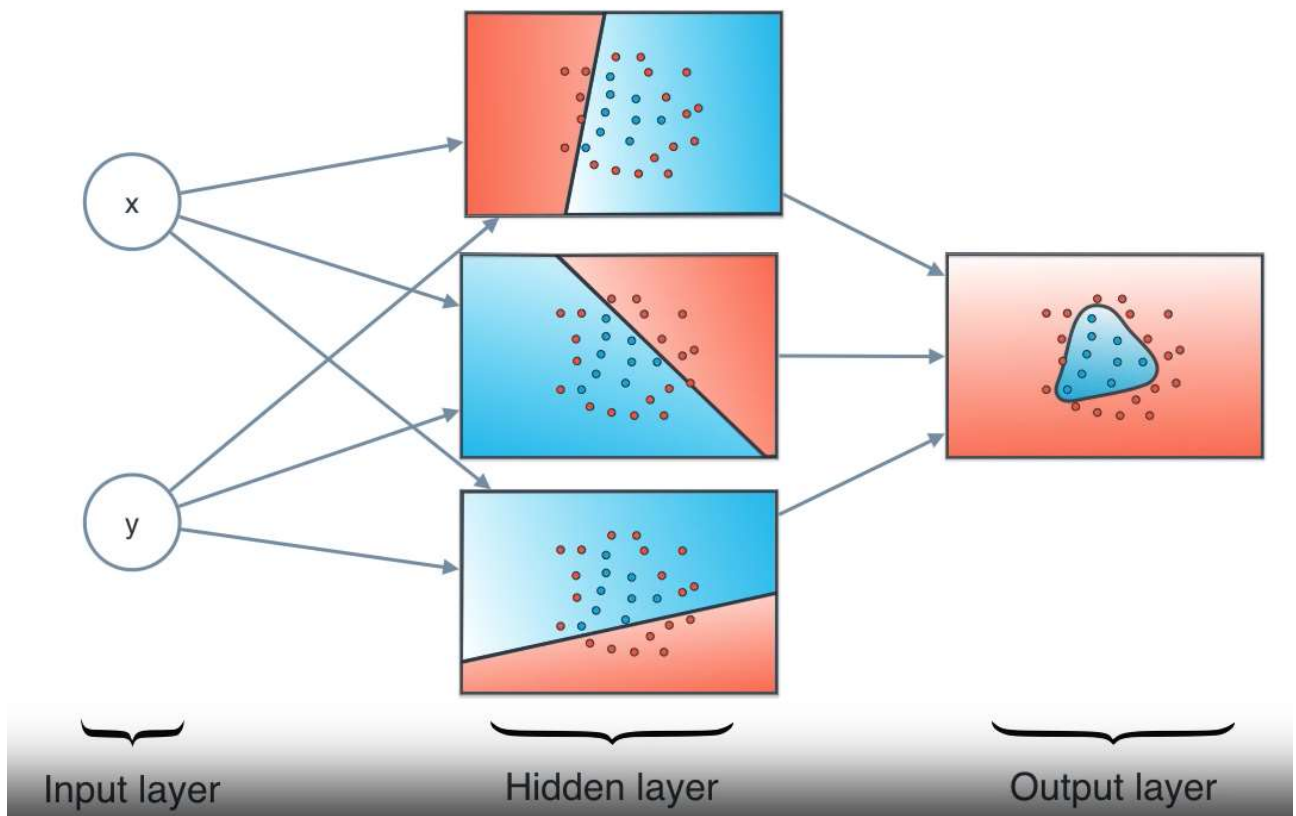


Neural Network

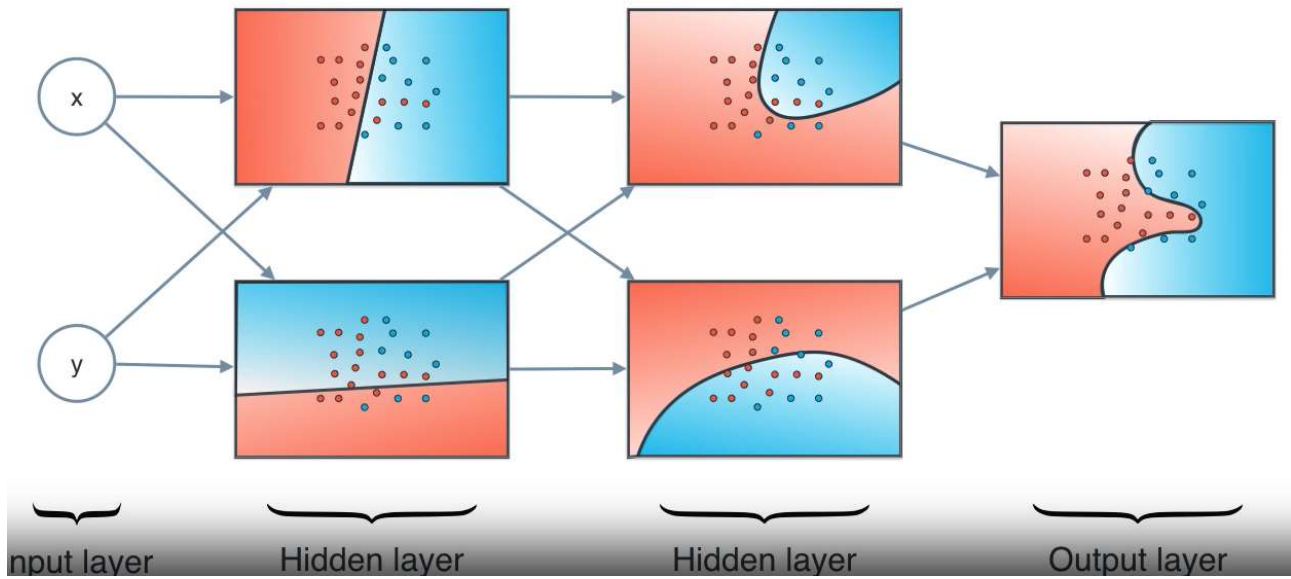


Neural Network

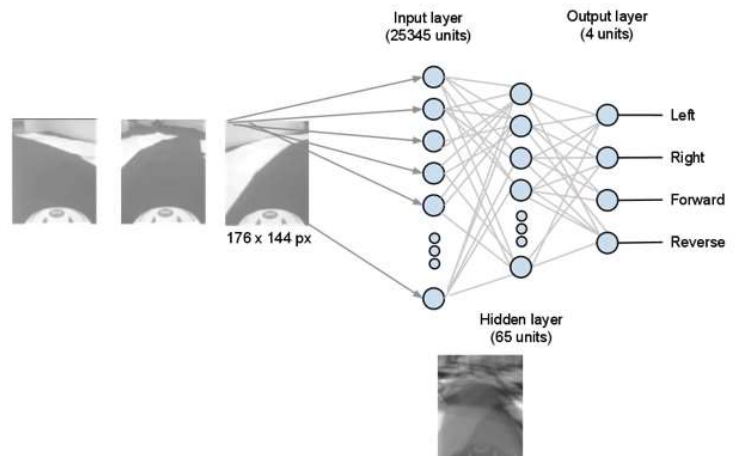




Deep Neural Network



Self Driving Car



CONVNET:

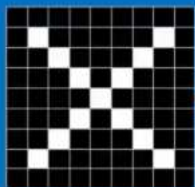
A toy ConvNet: X's and O's

Says whether a picture is of an X or an O

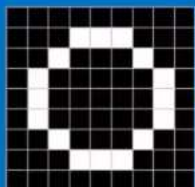
A two-dimensional
array of pixels



X or **O**

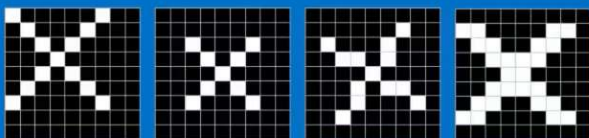


X



O

Trickier cases



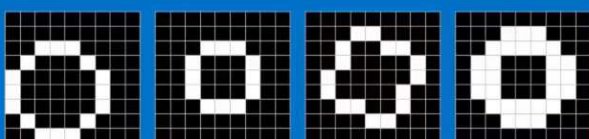
X

translation

scaling

rotation

weight



O

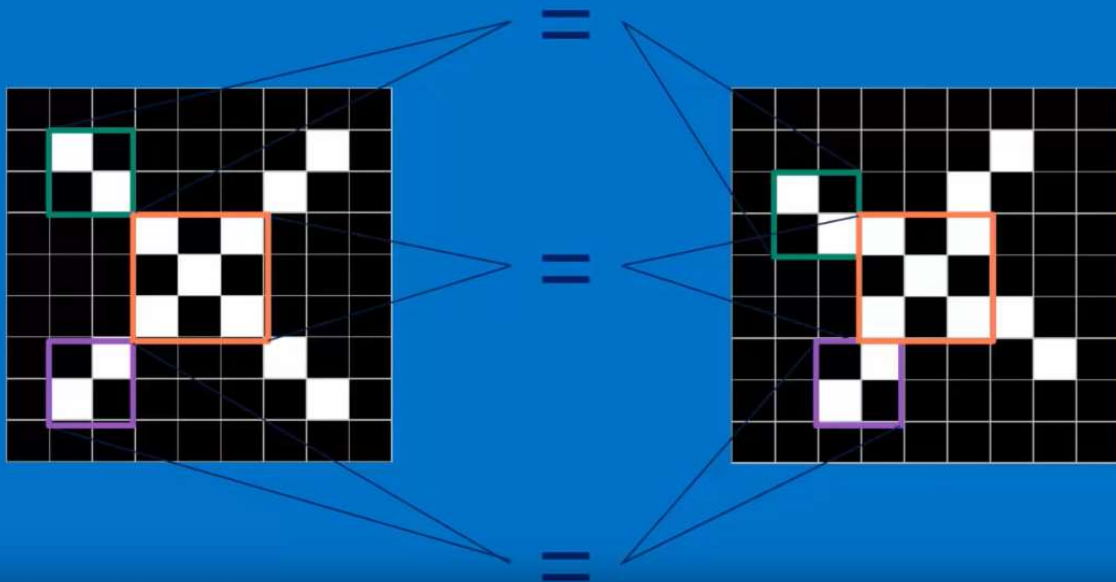
Computers are literal

-1	-1	-1	-1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
-1	-1	1	-1	-1	-1	1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	1	-1	-1	-1	1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
-1	-1	-1	-1	-1	-1	-1	-1	-1



-1	-1	-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	1	-1	-1
-1	1	-1	-1	-1	1	-1	-1	-1
-1	-1	1	1	-1	1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	1	-1	1	1	-1	-1
-1	-1	-1	1	-1	-1	-1	1	-1
-1	-1	1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1	-1	-1

ConvNets match pieces of the image



Features match pieces of the image

1	-1	-1
-1	1	-1
-1	-1	1

1	-1	1
-1	1	-1
1	-1	1

-1	-1	1
-1	1	-1
1	-1	-1

Filtering: The math behind the match

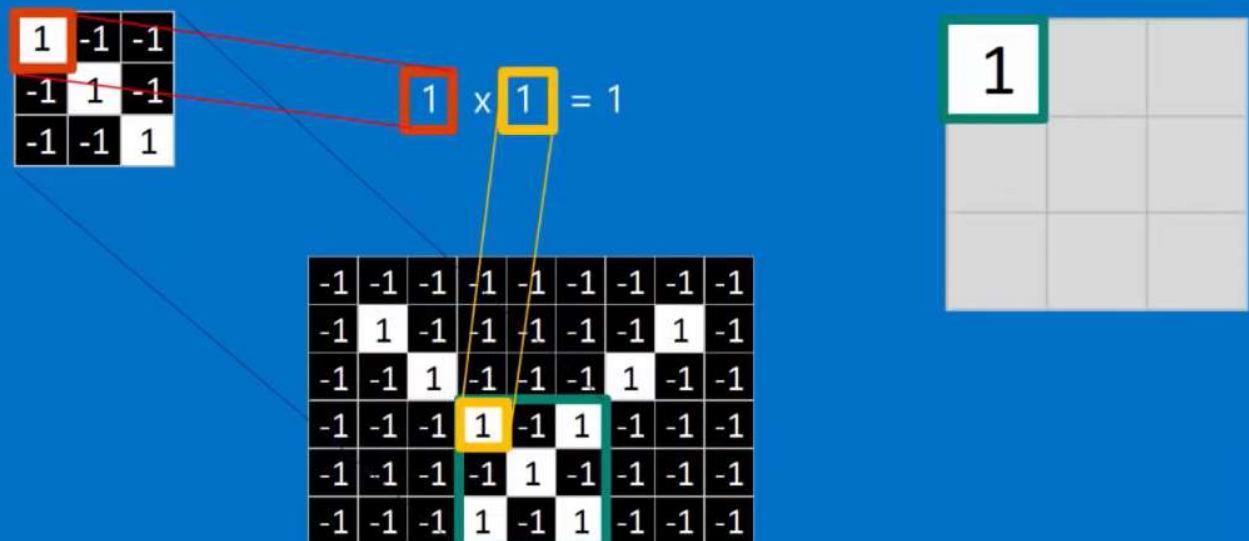
1	-1	-1
-1	1	-1
-1	-1	1

-1	-1	-1	-1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
-1	-1	1	-1	-1	-1	1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	1	-1	-1	1	1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
-1	-1	-1	-1	-1	-1	-1	-1	-1

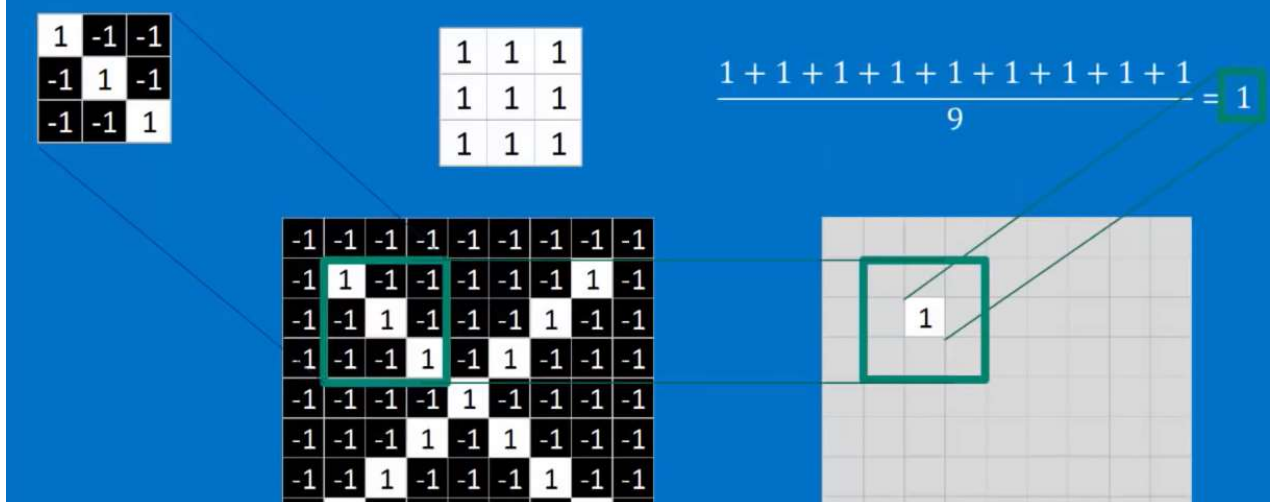
Filtering: The math behind the match

1. Line up the feature and the image patch.
2. Multiply each image pixel by the corresponding feature pixel.
3. Add them up.
4. Divide by the total number of pixels in the feature.

Filtering: The math behind the match



Filtering: The math behind the match



Convolution: Trying every possible match

1	-1	-1
-1	1	-1
-1	-1	1

-1	-1	-1	-1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
-1	-1	1	-1	-1	-1	1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	1	-1	-1	-1	1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
1	-1	-1	-1	-1	-1	-1	-1	1



0.77	-0.11	0.11	0.33	0.55	-0.11	0.33
-0.11	1.00	-0.11	0.33	-0.11	0.11	-0.11
0.11	-0.11	1.00	-0.33	0.11	-0.11	0.55
0.33	0.33	-0.33	0.55	-0.33	0.33	0.33
0.55	-0.11	0.11	-0.33	1.00	-0.11	0.11
-0.11	0.11	-0.11	0.33	-0.11	1.00	-0.11
0.33	-0.11	0.55	0.33	0.11	-0.11	0.77

-1	-1	-1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	-1	1
-1	-1	1	-1	-1	-1	1	-1
-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1
-1	-1	1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	1	-1
-1	-1	-1	-1	-1	-1	-1	-1



1	-1	-1
-1	1	-1
-1	-1	1



0.77	-0.11	0.11	0.33	0.55	-0.11	0.33
-0.11	1.00	-0.11	0.33	-0.11	0.11	-0.11
0.11	-0.11	1.00	-0.33	0.11	-0.11	0.55
0.33	0.33	-0.33	0.55	-0.33	0.33	0.33
0.55	-0.11	0.11	-0.33	1.00	-0.11	0.11
-0.11	0.11	-0.11	0.33	-0.11	1.00	-0.11
0.33	-0.11	0.55	0.33	0.11	-0.11	0.77

-1	-1	-1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	-1	1
-1	-1	1	-1	-1	-1	1	-1
-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1
-1	-1	1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	1	-1
-1	-1	-1	-1	-1	-1	-1	-1



1	-1	1
-1	1	-1
1	-1	1



0.33	-0.55	0.11	-0.11	0.11	-0.55	0.33
-0.55	0.55	-0.55	0.33	-0.55	0.55	-0.55
0.11	-0.55	0.55	-0.77	0.55	-0.55	0.11
-0.11	0.33	-0.77	1.00	-0.77	0.33	-0.11
0.11	-0.55	0.55	-0.77	0.55	-0.55	0.11
-0.55	0.55	-0.55	0.33	-0.55	0.55	-0.55
0.33	-0.55	0.11	-0.11	0.11	-0.55	0.33

-1	-1	-1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	-1	1
-1	-1	1	-1	-1	-1	1	-1
-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1
-1	-1	1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	1	-1
-1	-1	-1	-1	-1	-1	-1	-1

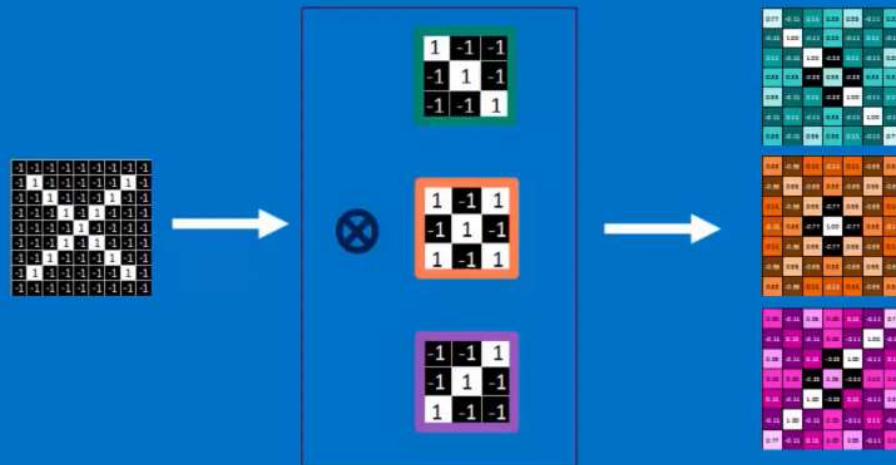


-1	-1	1
-1	1	-1
1	-1	-1



0.33	-0.11	0.55	0.33	0.11	-0.11	0.77
-0.11	0.11	-0.11	0.33	-0.11	1.00	-0.11
0.55	-0.11	0.11	-0.33	1.00	-0.11	0.11
0.33	0.33	-0.33	0.55	-0.33	0.33	0.33
0.11	-0.11	1.00	-0.33	0.11	-0.11	0.55
-0.11	1.00	-0.11	0.33	-0.11	0.11	-0.11
0.77	-0.11	0.11	0.33	0.55	-0.11	0.33

One image becomes a stack of filtered images



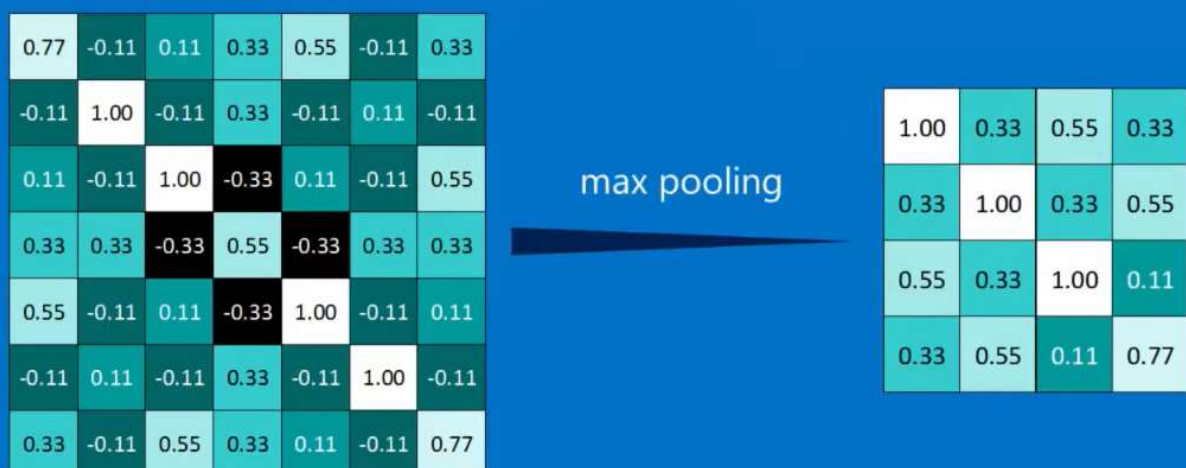
Pooling: Shrinking the image stack

1. Pick a window size (usually 2 or 3).
2. Pick a stride (usually 2).
3. Walk your window across your filtered images.
4. From each window, take the maximum value.

Pooling

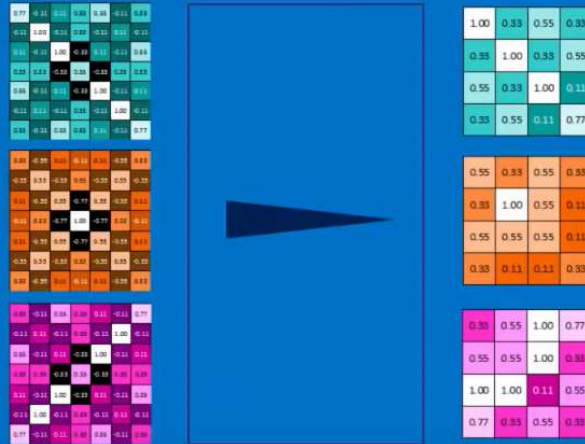


Pooling



Pooling layer

A stack of images becomes a stack of smaller images.



Normalization

Keep the math from breaking by tweaking each of the values just a bit.

Change everything negative to zero.

Rectified Linear Units (ReLUs)



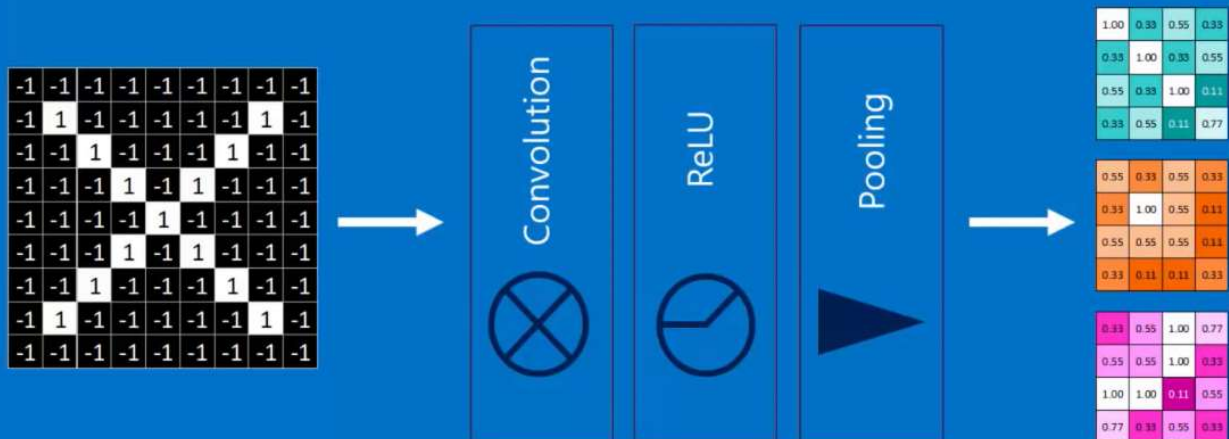
ReLU layer

A stack of images becomes a stack of images with no negative values.



Layers get stacked

The output of one becomes the input of the next.



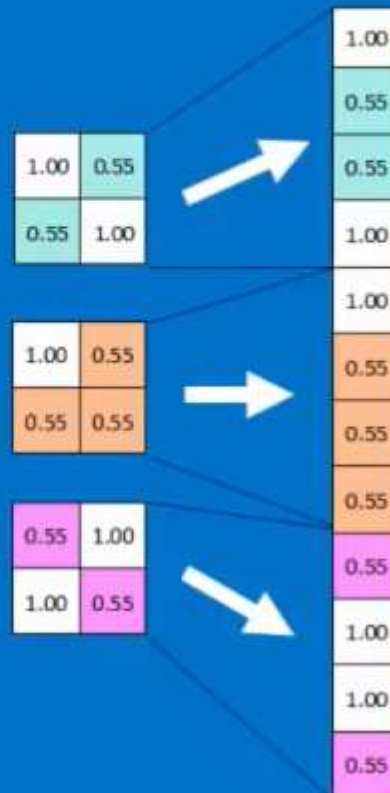
Deep stacking

Layers can be repeated several (or many) times.



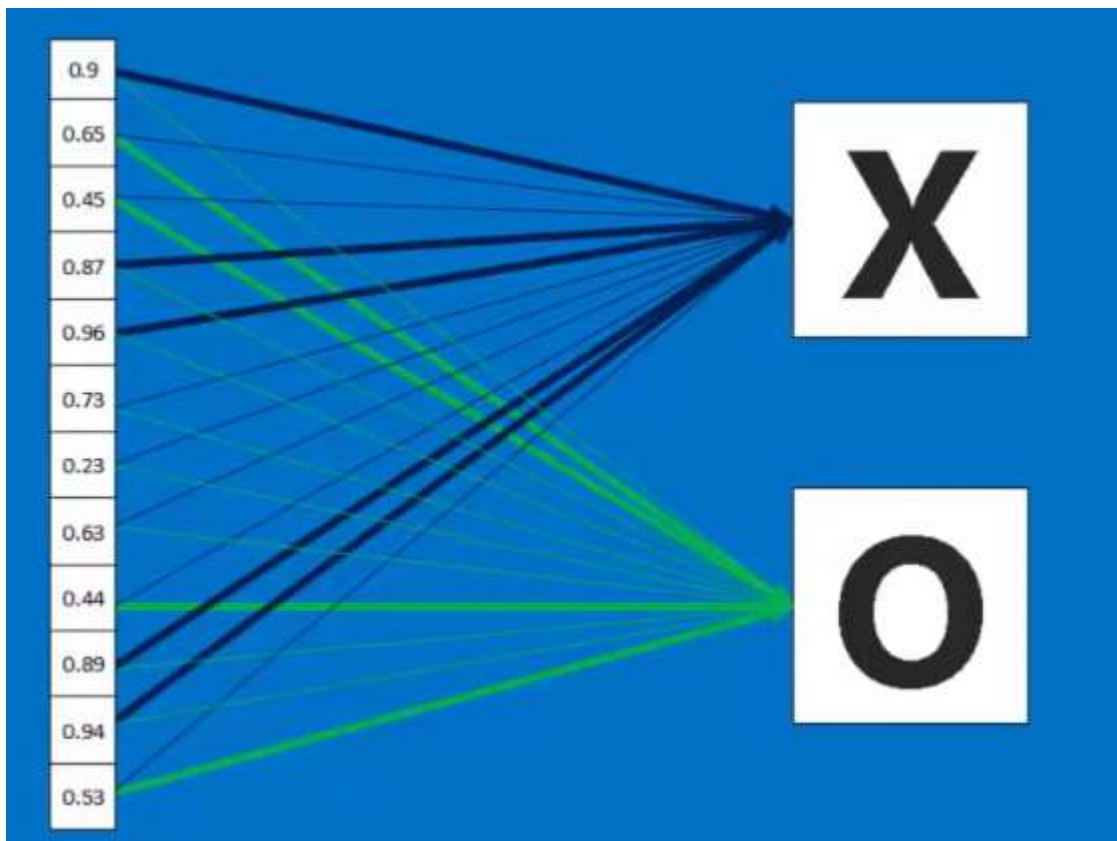
Fully connected layer

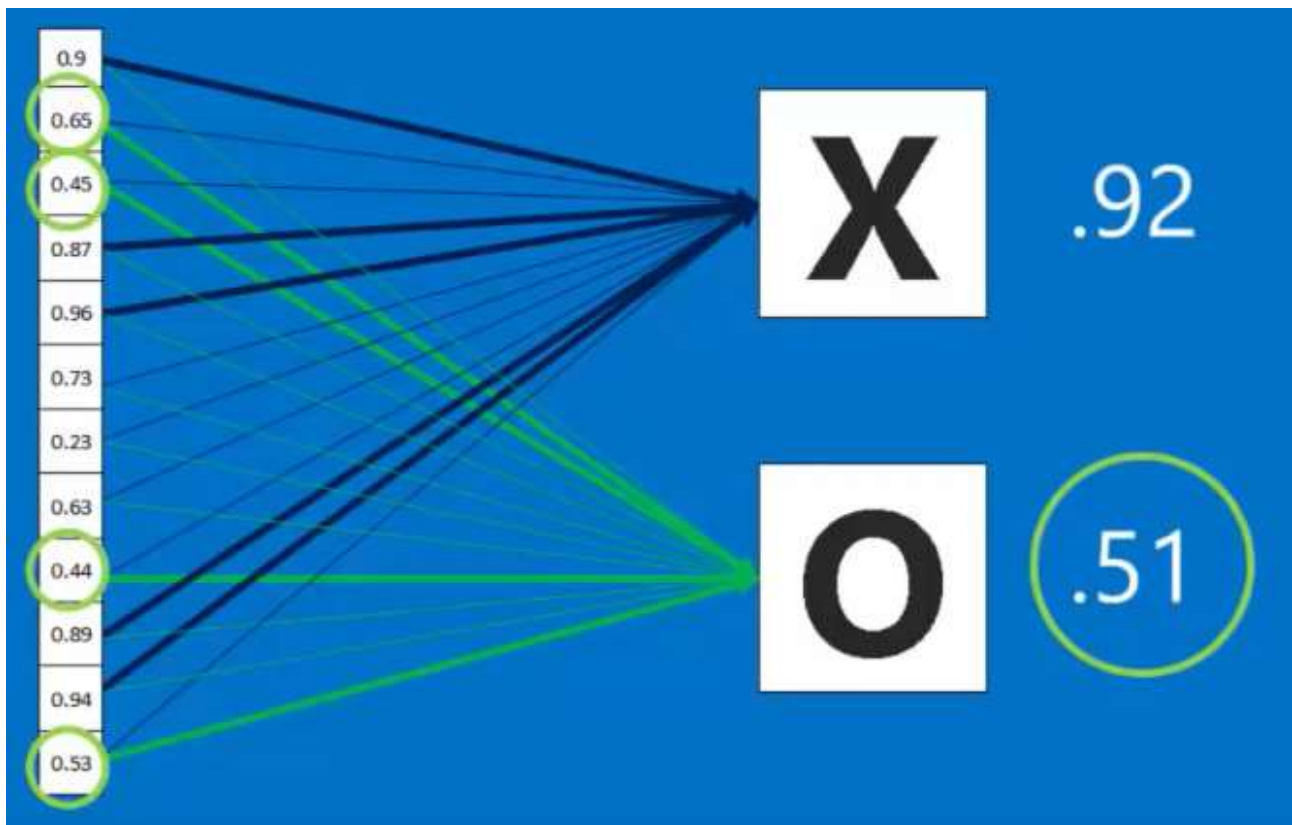
Every value gets a vote



Fully connected layer

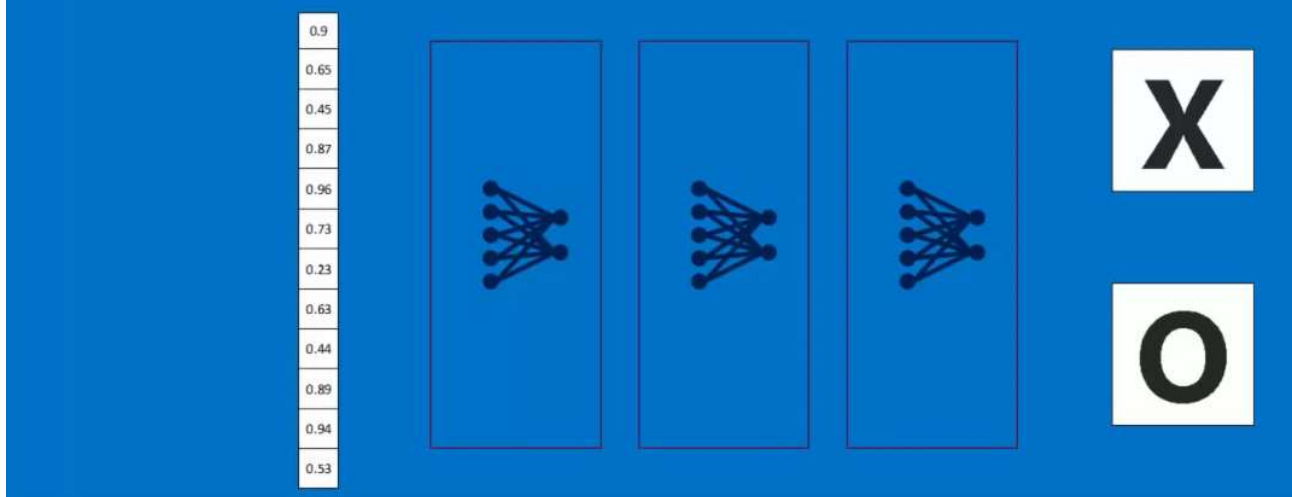
Vote depends on how strongly a value predicts X or O





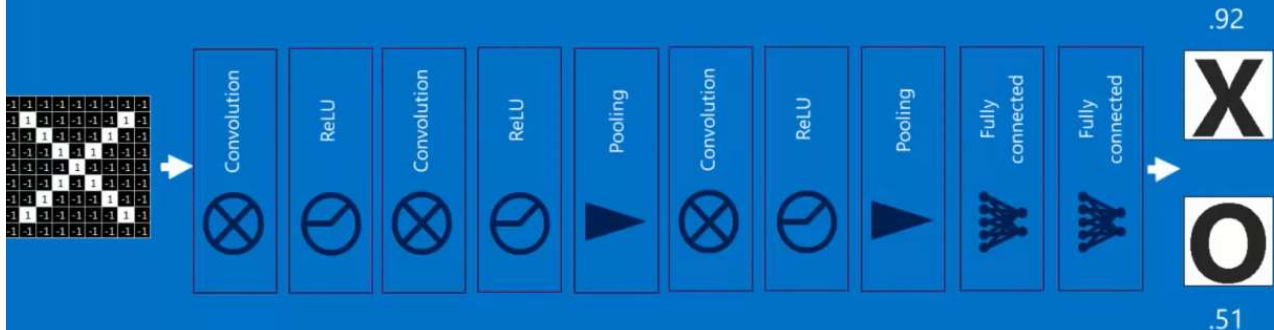
Fully connected layer

These can also be stacked.



Putting it all together

A set of pixels becomes a set of votes.



Learning

Q: Where do all the magic numbers come from?

Features in convolutional layers

Voting weights in fully connected layers

A: Backpropagation

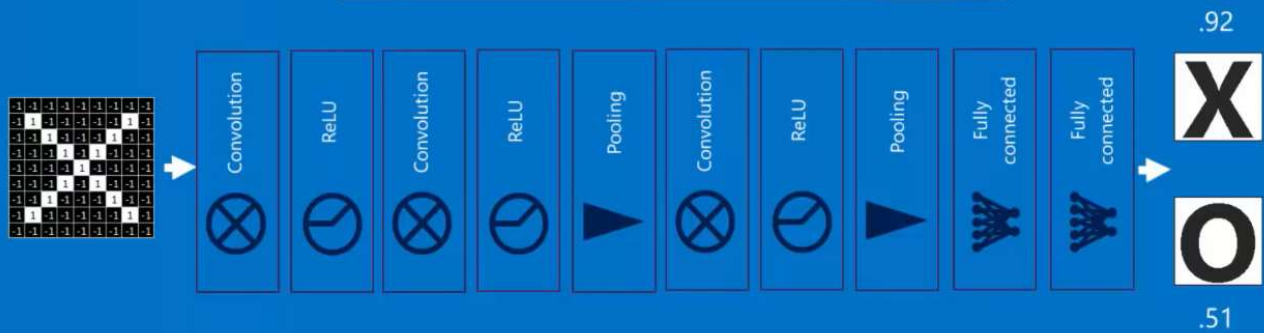
Backprop

Error = right answer – actual answer



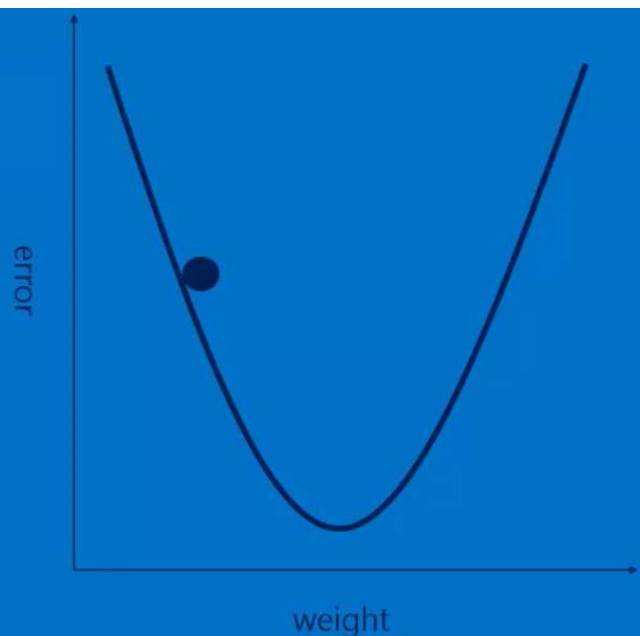
Backprop

	Right answer	Actual answer	Error
X	1	0.92	0.08
O	0	0.51	0.49
		Total	0.57



Gradient descent

For each feature pixel and voting weight, adjust it up and down a bit and see how the error changes.



Hyperparameters (knobs)

Convolution

- Number of features

- Size of features

Pooling

- Window size

- Window stride

Fully Connected

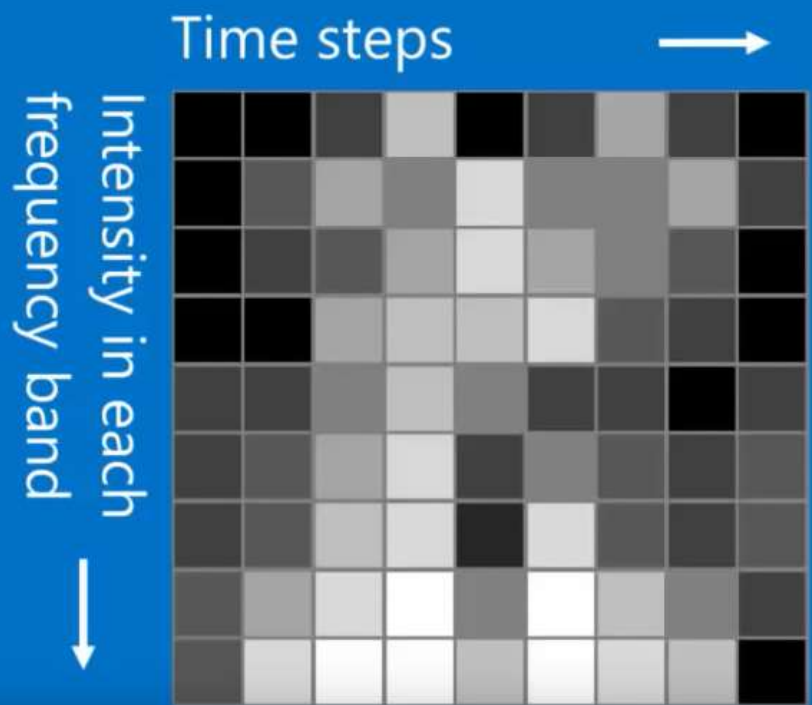
- Number of neurons

Architecture

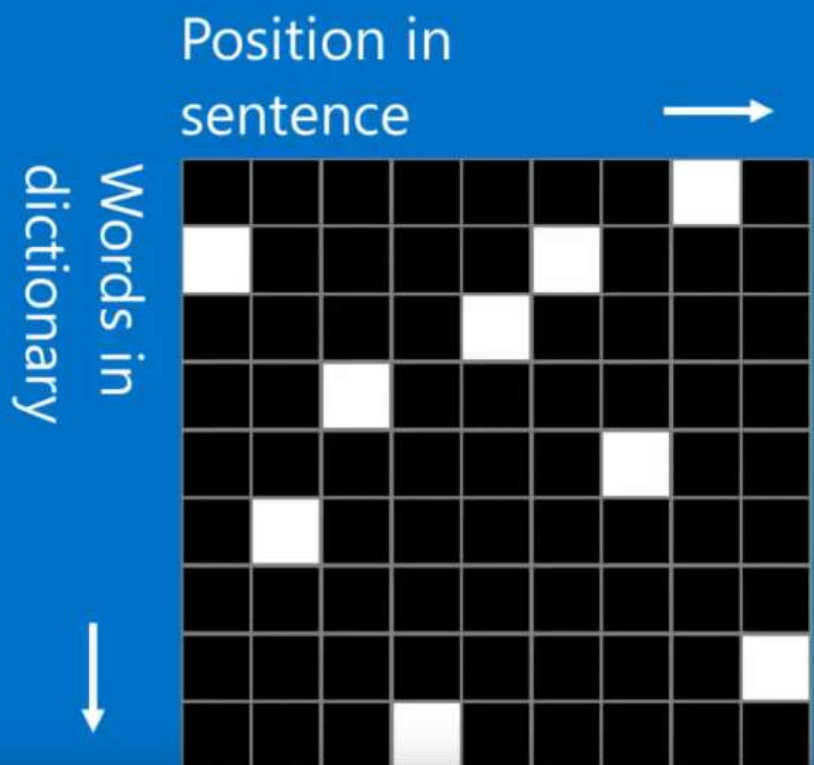
How many of each type of layer?

In what order?

Sound



Text



Some ConvNet/DNN toolkits

Caffe (Berkeley Vision and Learning Center)

CNTK (Microsoft)

Deeplearning4j (Skymind)

TensorFlow (Google)

Theano (University of Montreal + broad community)

Torch (Ronan Collobert)

Many others