



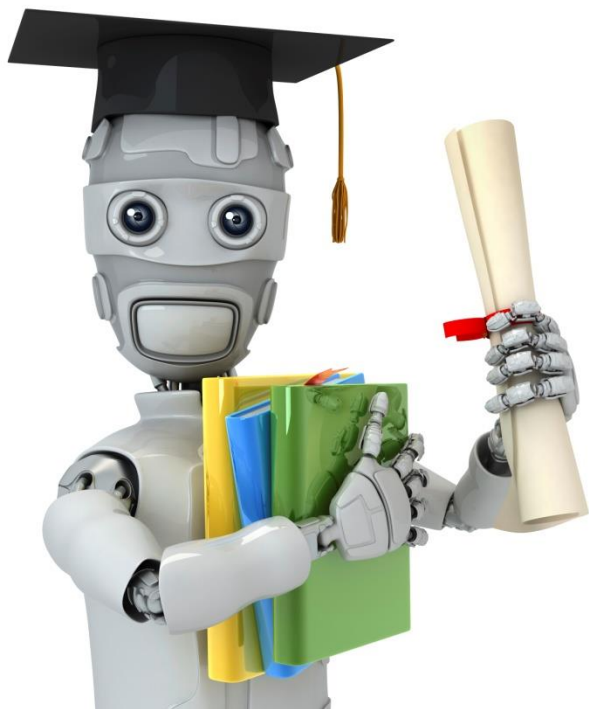
CSE 4621

Machine Learning

Lecture 7

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Islamic University of Technology (IUT)





Neural Network

Introduction

Machine Learning

Source & Special Thanks to (Coursera) Machine Learning / NN&DL Courses

Computer Vision: Car detection



Cars

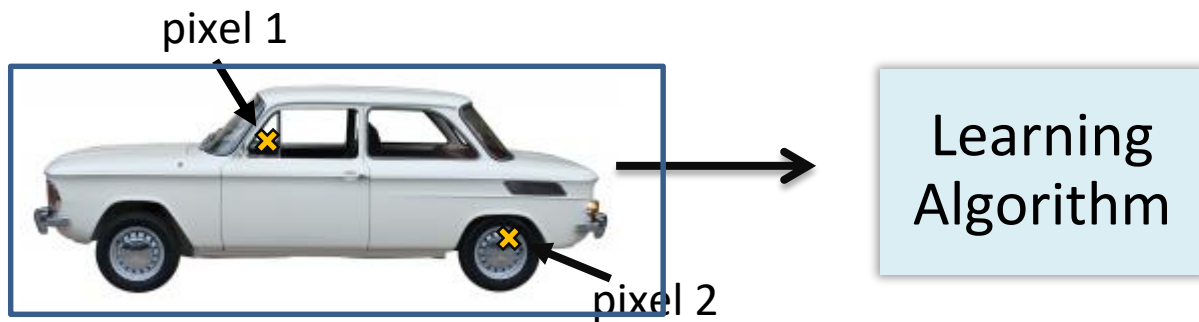


Not a car

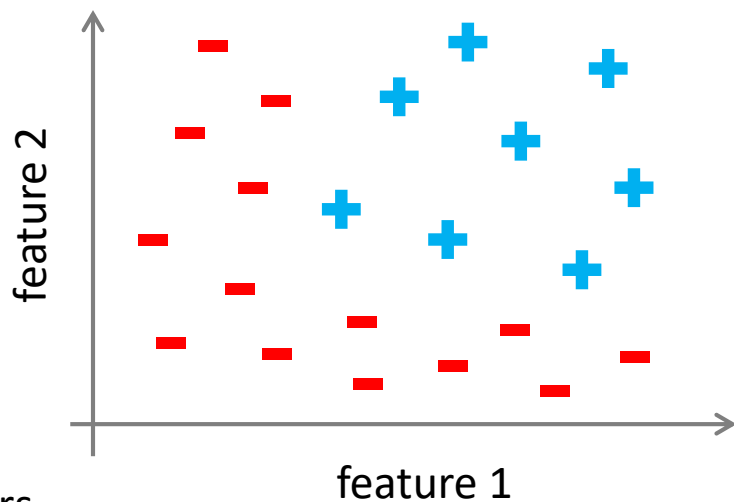
Testing:



What is this?



50 x 50 pixel images \rightarrow 2500 pixels
 $n = 2500$ (7500 if RGB)



+ Cars
- "Non"-Cars

$$x = \begin{bmatrix} \text{pixel 1 intensity} \\ \text{pixel 2 intensity} \\ \vdots \\ \text{pixel 2500 intensity} \end{bmatrix}$$

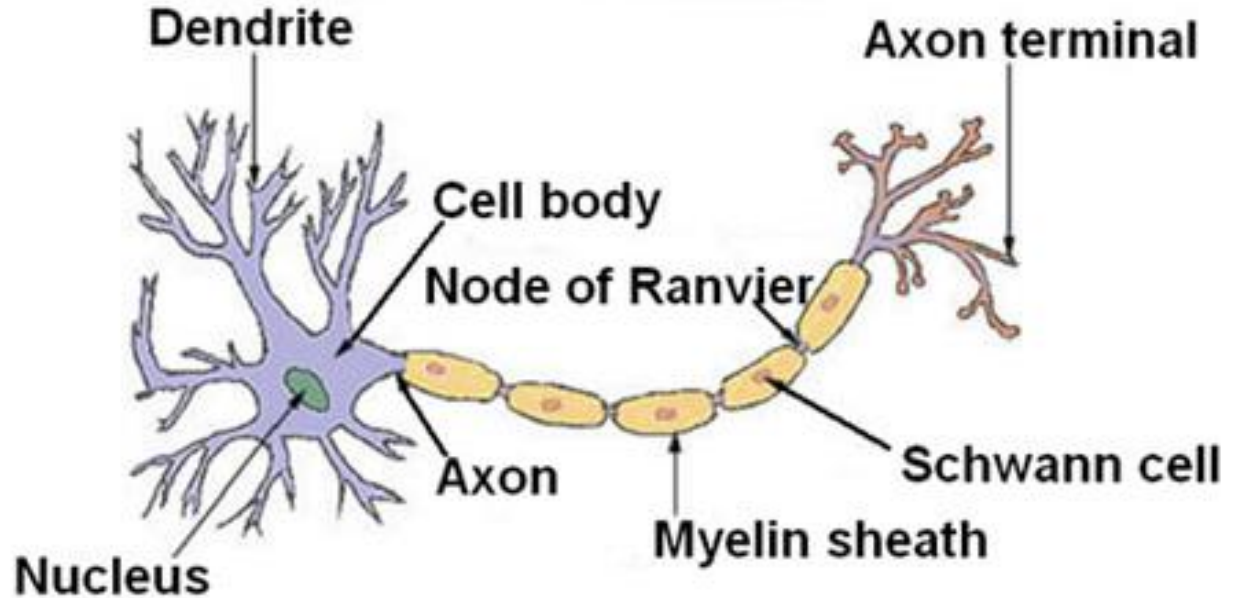
Quadratic features ($x_i \times x_j$): ≈ 3 million features $O(n^2)$

Neural Networks (NN)

- Origins: Algorithms that try to mimic the brain.
- Was very widely used in 80s and early 90s;
 - popularity diminished in late 90s.
- Recent resurgence: State-of-the-art technique for many applications & due to high computing devices.
- NN is another non-linear classifier
 - Decision boundary is non-linear

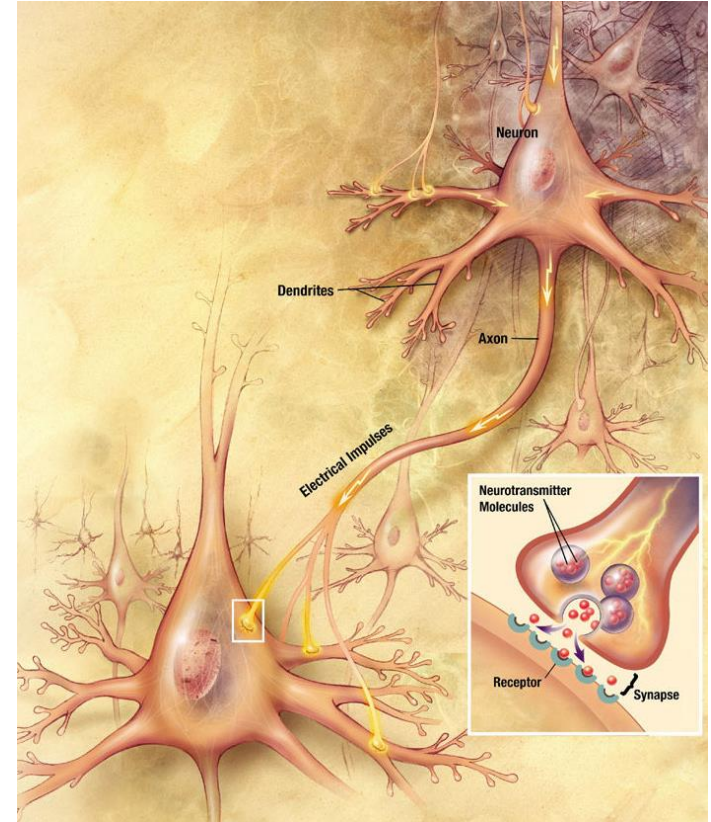
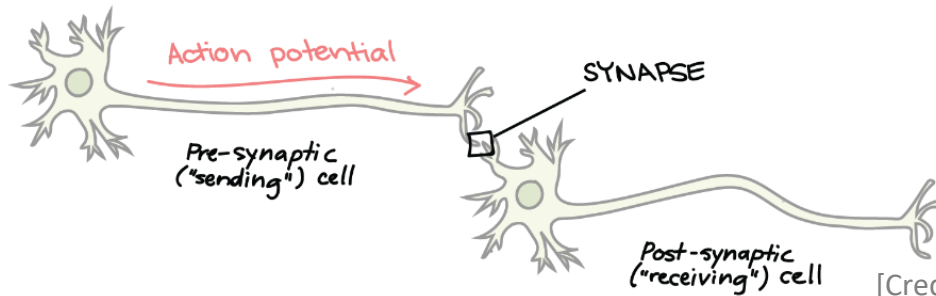
Neuron in the brain

- Biological Neurons are the core components of the human brain.
- A neuron consists of a cell body, dendrites, and an axon.



Neurons in the brain

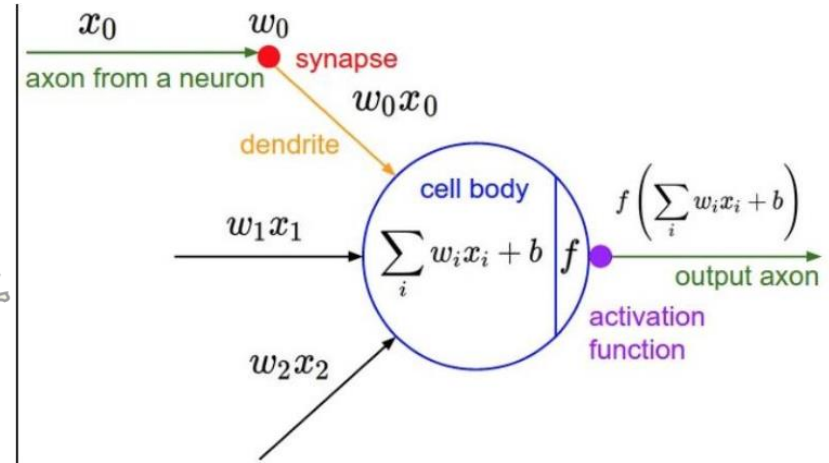
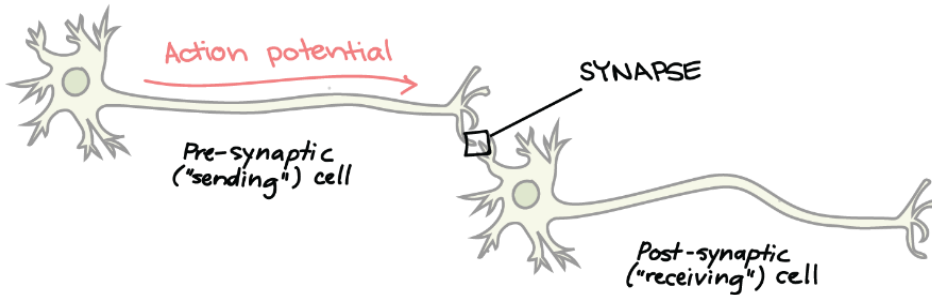
- Neurons process and transmit information to other neurons by emitting electrical signals.
- Each neuron receives input signals from its dendrites and produces output signals along its axon.
- The axon branches out and connects via synapses to dendrites of other neurons.



[Credit: US National Institutes of Health, National Institute on Aging]

Biological Neuron VS. Artificial Neuron

- Artificial neurons are inspired by biological neurons
- An artificial neuron has a finite number of inputs with weights associated to them, and an activation function (also called transfer function).
- The output of the neuron is the result of the activation function applied to the weighted sum of inputs.
- Artificial neurons are connected with each others to form artificial neural networks.

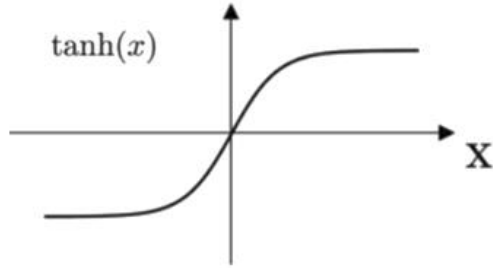


Activation Function

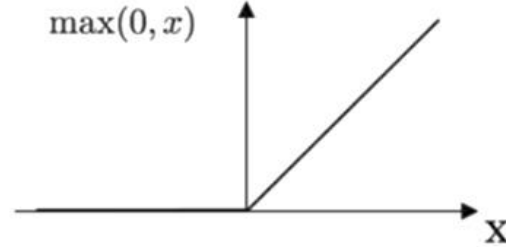
- Activation functions transform the weighted sum of inputs that goes into the artificial neurons.
- These functions should be non-linear to encode complex patterns of the data.
- The most popular activation functions are
 - Logistic
 - Tanh
 - Rectified Linear Unit (RELU)
 - Leaky RELU
 - Exponential Linear Unit (ELU)

Activation Function

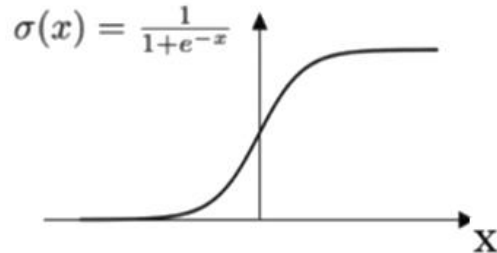
Tanh



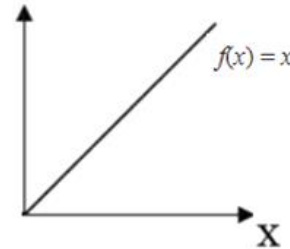
ReLU



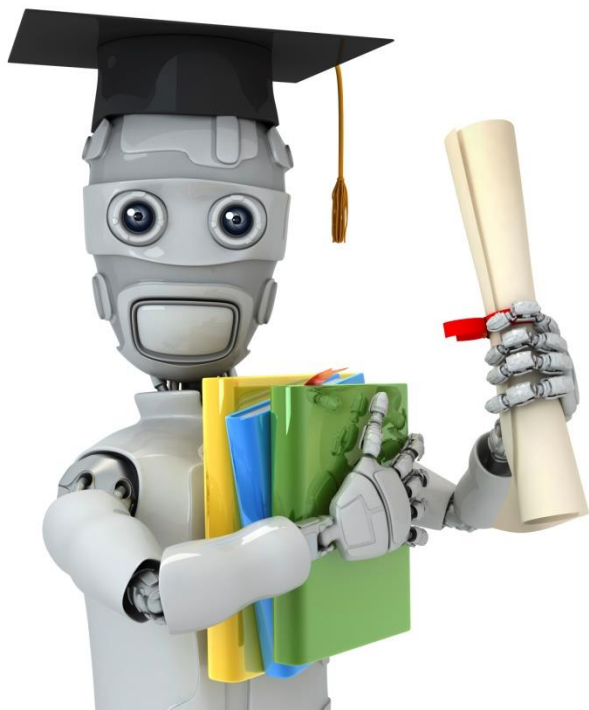
Sigmoid



Linear



Have a look for more: https://en.wikipedia.org/wiki/Activation_function



Machine Learning

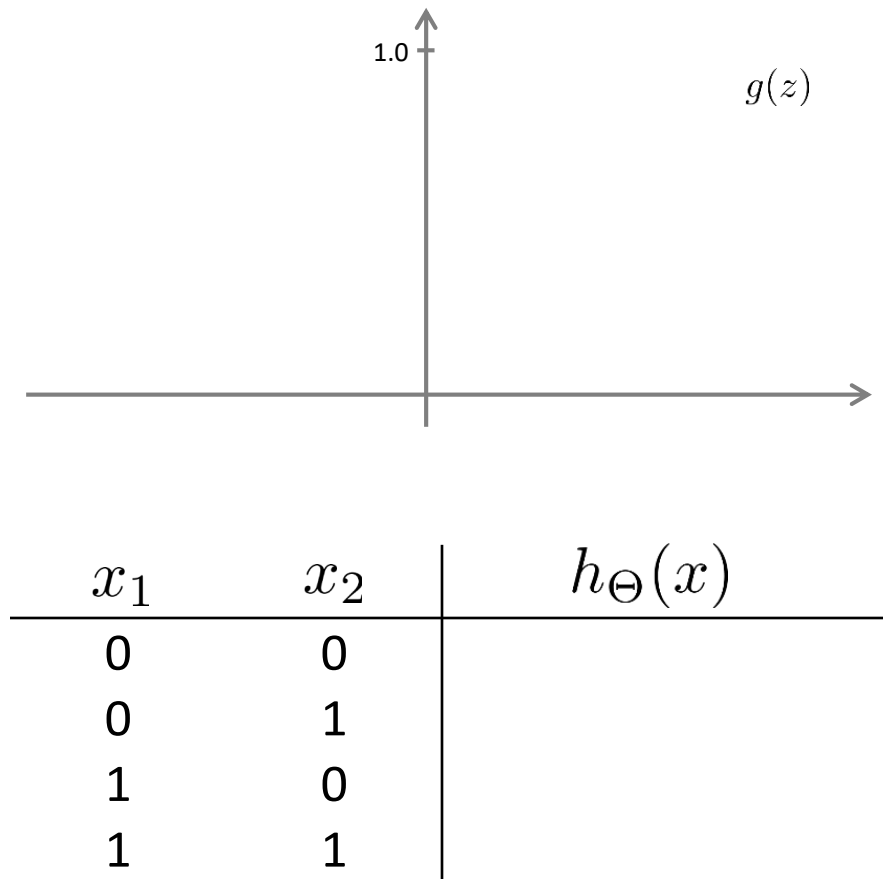
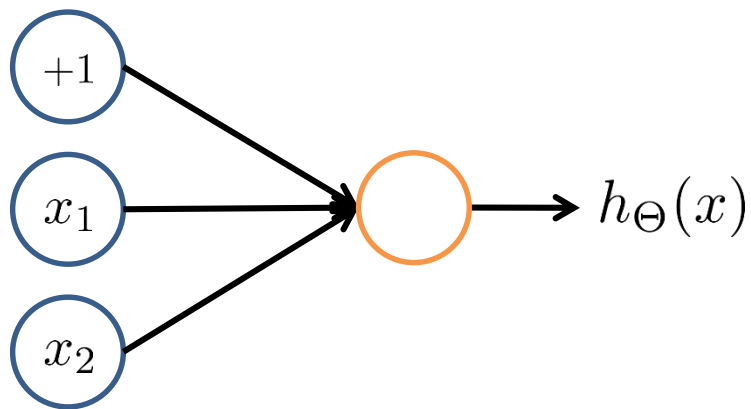
Neural Networks: Representation

Examples and intuitions II

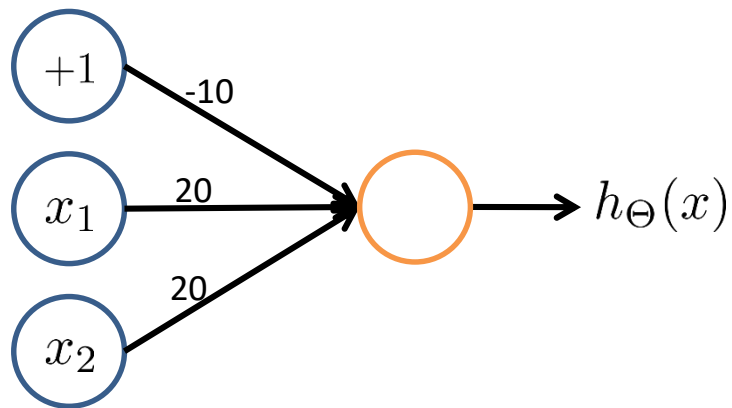
Simple example: AND

$$x_1, x_2 \in \{0, 1\}$$

$$y = x_1 \text{ AND } x_2$$



Example: OR function

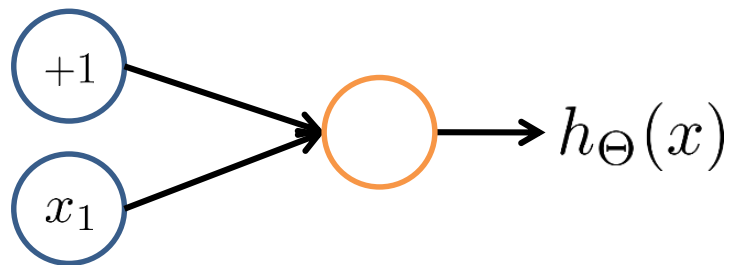


x_1	x_2	$h_{\Theta}(x)$
0	0	
0	1	
1	0	
1	1	

x_1 AND x_2

x_1 OR x_2

Negation:



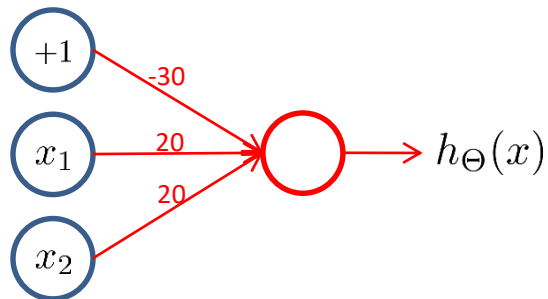
$$h_{\Theta}(x) = g(10 - 20x_1)$$

x_1	$h_{\Theta}(x)$
0	
1	

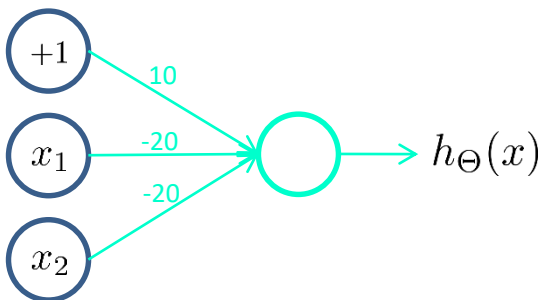
(NOT x_1) AND (NOT x_2)

nor

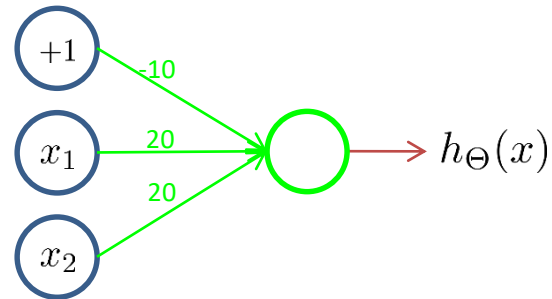
Putting it together: x_1 XNOR x_2



x_1 AND x_2



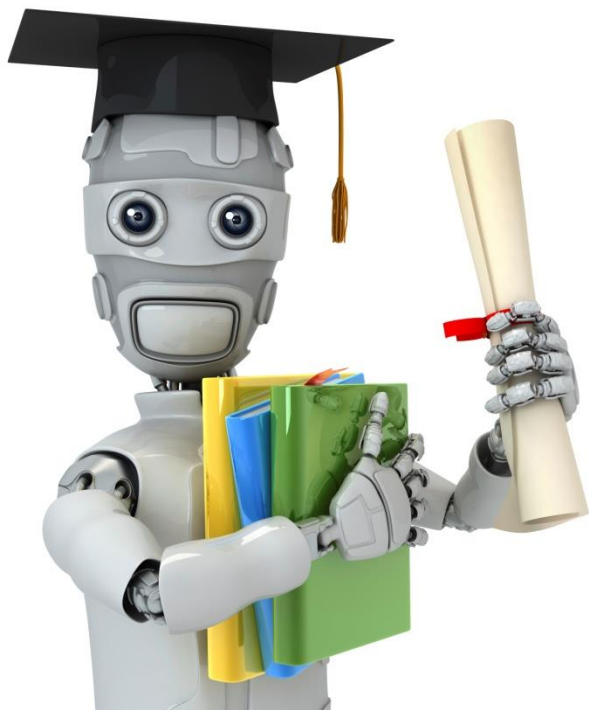
(NOT x_1) AND (NOT x_2)



x_1 OR x_2



x_1	x_2	$a_1^{(2)}$	$a_2^{(2)}$	$h_{\Theta}(x)$
0	0			
0	1			
1	0			
1	1			

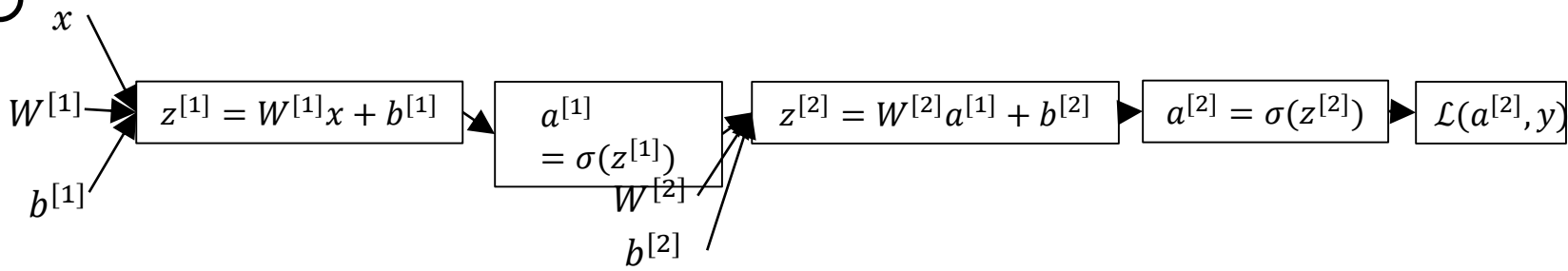
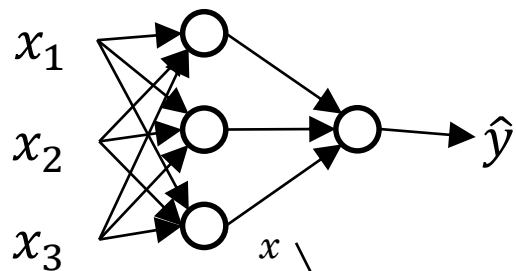
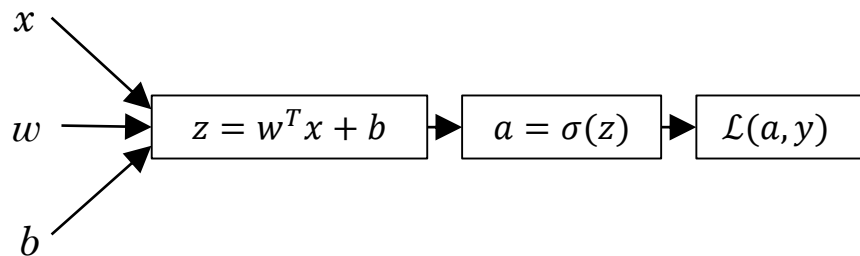
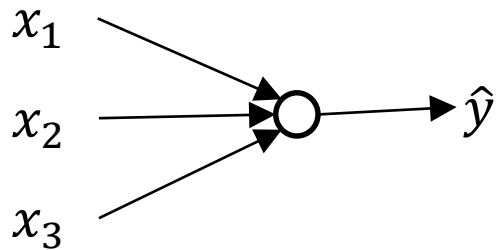


Machine Learning

Neural Networks: Representation

Two-class classification

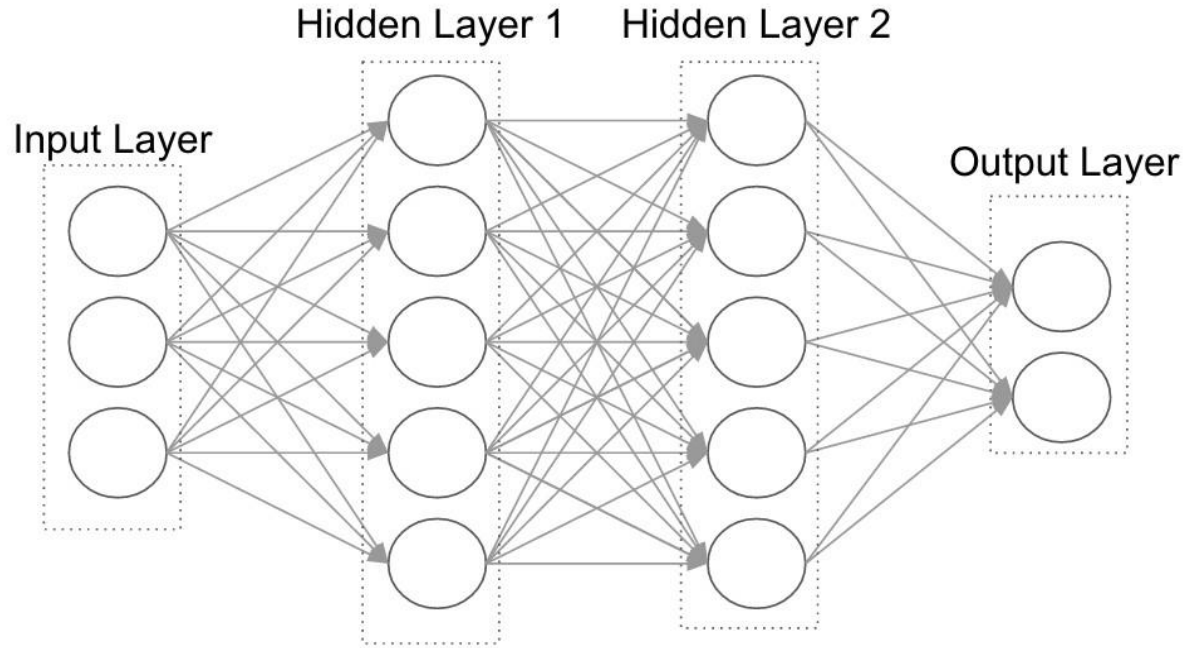
Define a Neural Network



Feed Forward NN

- Feed forward Neural Networks (FFN) are the simplest form of Artificial Neural Networks (ANN).
- These networks have 3 types of layers:
 - Input layer,
 - hidden layer and
 - output layer.
- In FFN, data moves from the input layer through the hidden nodes (if any) and to the output nodes.
- “Fully-connected” means that each node is connected to all the nodes in the next layer.
- The number of hidden layers and their size are the parameters. The larger and deeper the hidden layers, the more complex patterns we can model in theory.

Multi-layer Perceptron (MLP)



Feedforward neural network with 2 hidden layers

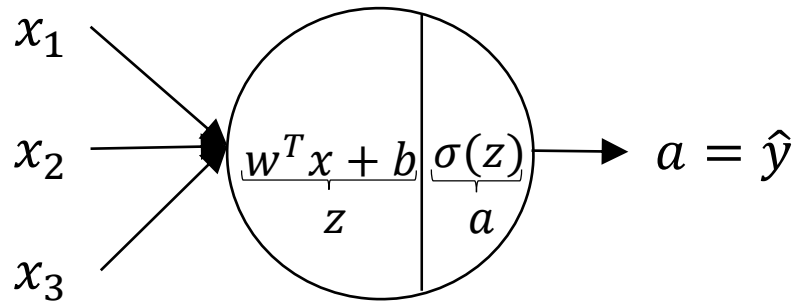


deeplearning.ai

One hidden layer Neural Network

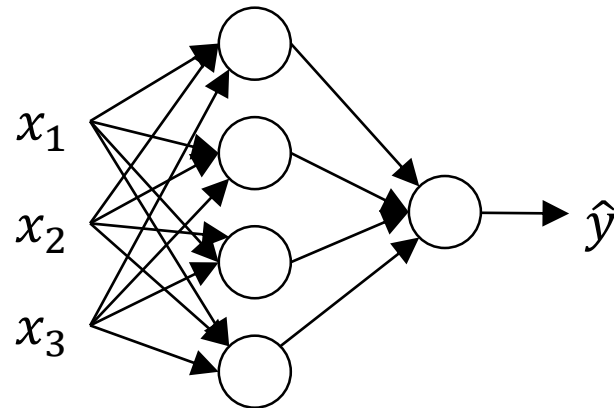
Computing a
Neural Network's
Output

Neural Network Representation

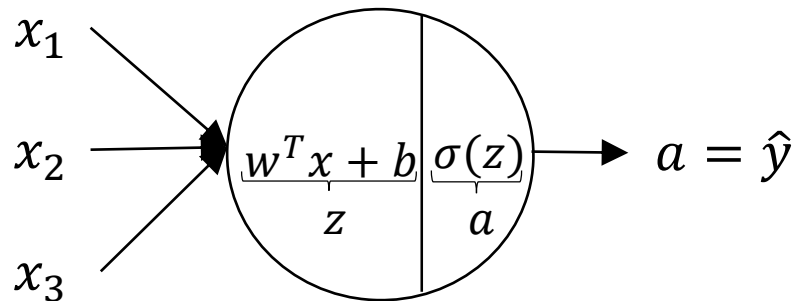


$$z = w^T x + b$$

$$a = \sigma(z)$$

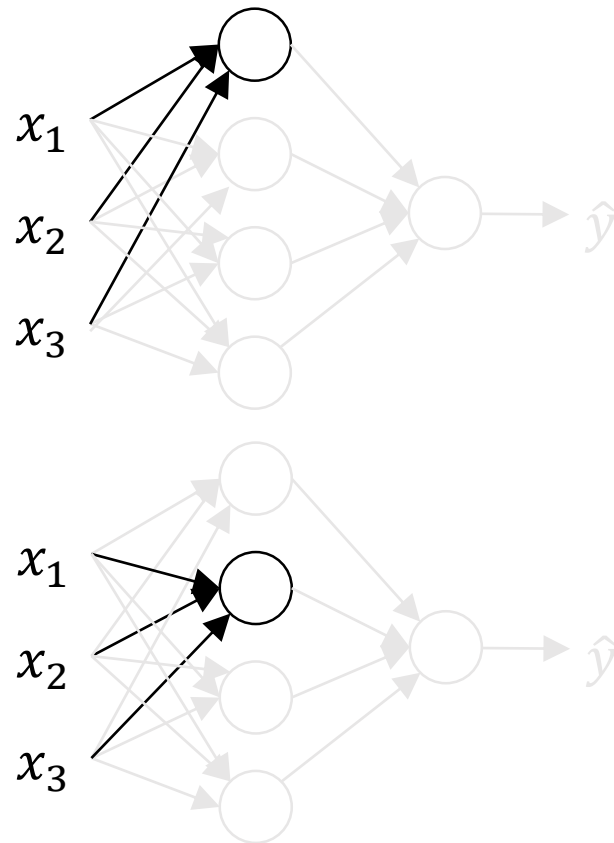


Neural Network Representation

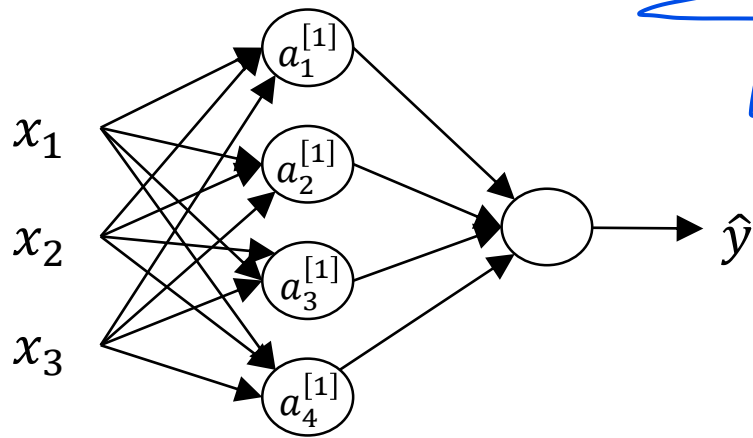


$$z = w^T x + b$$

$$a = \sigma(z)$$



Neural Network Representation



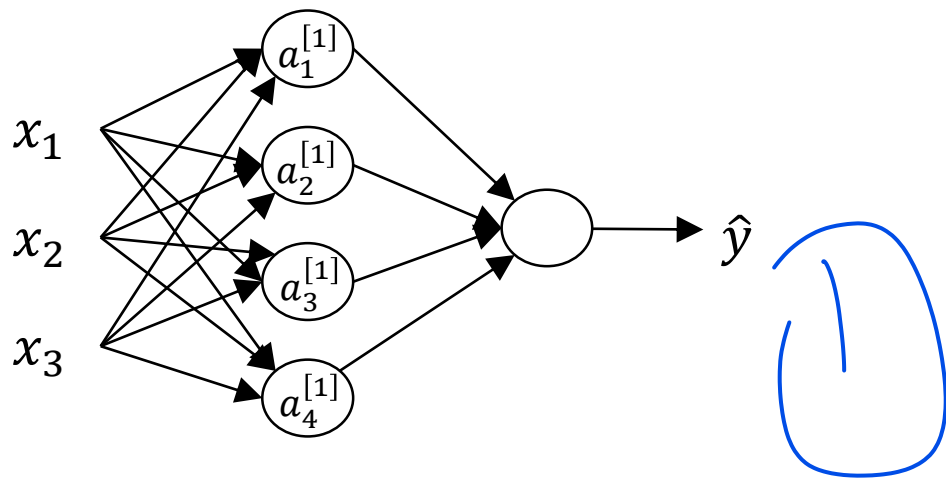
$$z_1^{[1]} = w_1^{[1]T} x + b_1^{[1]}, a_1^{[1]} = \sigma(z_1^{[1]})$$

$$z_2^{[1]} = w_2^{[1]T} x + b_2^{[1]}, a_2^{[1]} = \sigma(z_2^{[1]})$$

$$z_3^{[1]} = w_3^{[1]T} x + b_3^{[1]}, a_3^{[1]} = \sigma(z_3^{[1]})$$

$$z_4^{[1]} = w_4^{[1]T} x + b_4^{[1]}, a_4^{[1]} = \sigma(z_4^{[1]})$$

Neural Network Representation



Given input x :

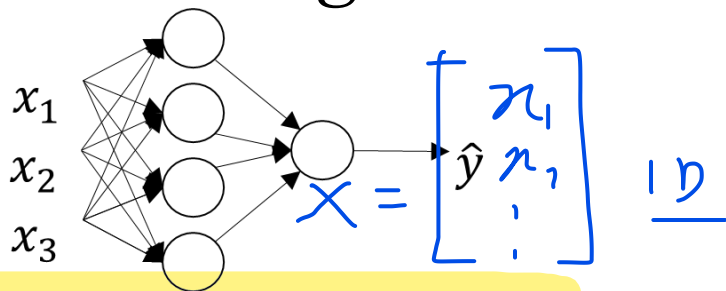
$$z^{[1]} = W^{[1]}x + b^{[1]}$$

$$a^{[1]} = \sigma(z^{[1]})$$

$$z^{[2]} = W^{[2]}a^{[1]} + b^{[2]}$$

$$a^{[2]} = \sigma(z^{[2]})$$

Vectorizing across multiple examples



$$X = \begin{bmatrix} x^{(1)} & x^{(2)} & \dots & x^{(m)} \end{bmatrix}$$

Handwritten blue annotations: $x^{(1)}, x^{(2)}, \dots, x^{(m)}$ are each enclosed in a blue box, and a blue arrow points from the first box to the second.

$$A^{[1]} = \begin{bmatrix} a^{1} & a^{[1](2)} & \dots & a^{[1](m)} \end{bmatrix}$$

Handwritten blue annotations: $a^{1}, a^{[1](2)}, \dots, a^{[1](m)}$ are each underlined in blue.

for $i = 1$ to m

$$z^{[1](i)} = W^{[1]}x^{(i)} + b^{[1]}$$

$$a^{[1](i)} = \sigma(z^{[1](i)})$$

$$z^{[2](i)} = W^{[2]}a^{[1](i)} + b^{[2]}$$

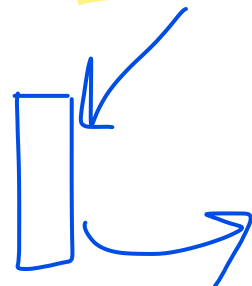
$$a^{[2](i)} = \sigma(z^{[2](i)})$$

$$Z^{[1]} = W^{[1]}X + b^{[1]}$$

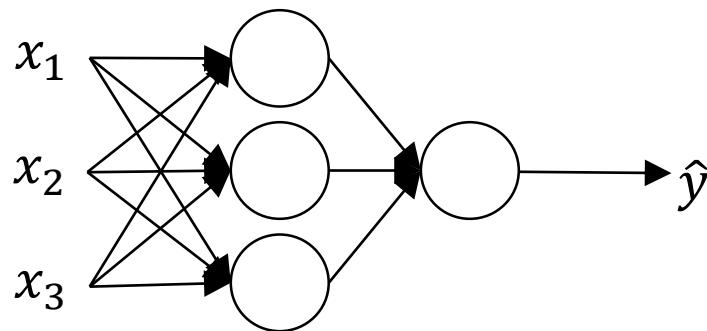
$$A^{[1]} = \sigma(Z^{[1]})$$

$$Z^{[2]} = W^{[2]}A^{[1]} + b^{[2]}$$

$$A^{[2]} = \sigma(Z^{[2]})$$



Why Activation function?



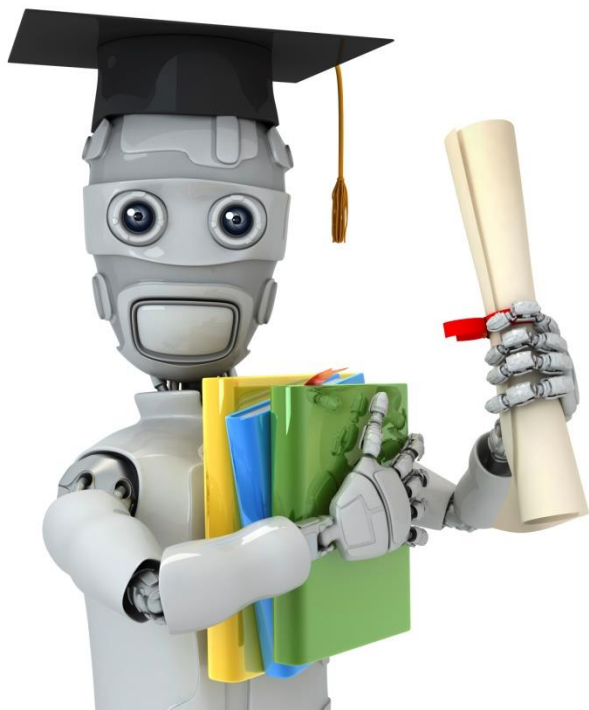
Given x :

$$z^{[1]} = W^{[1]}x + b^{[1]}$$

$$a^{[1]} = g^{[1]}(z^{[1]})$$

$$z^{[2]} = W^{[2]}a^{[1]} + b^{[2]}$$

$$a^{[2]} = g^{[2]}(z^{[2]})$$



Machine Learning

Neural Networks: Representation

Multi-class classification

Multiple output units: One-vs-all.



Pedestrian



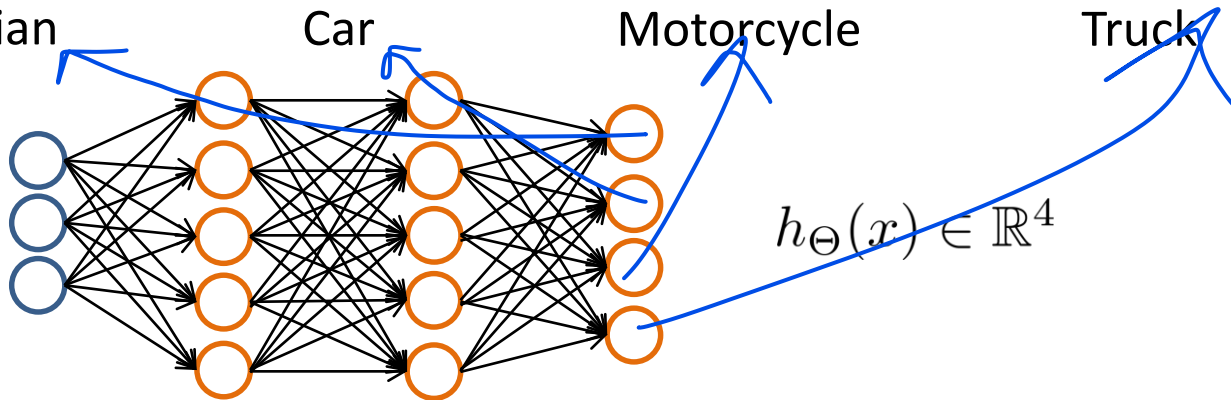
Car



Motorcycle



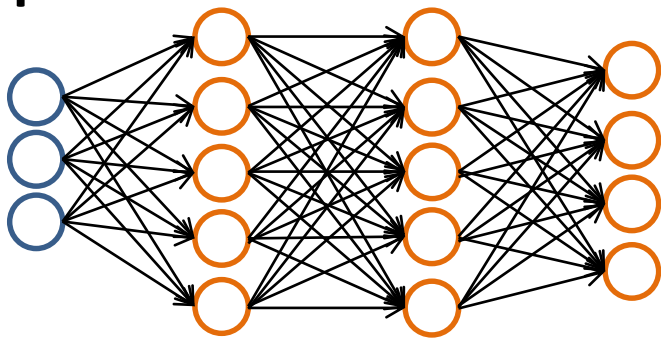
Truck



Want $h_{\Theta}(x) \approx \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$, $h_{\Theta}(x) \approx \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$, $h_{\Theta}(x) \approx \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$, etc.

when pedestrian when car when motorcycle

Multiple output units: One-vs-all.



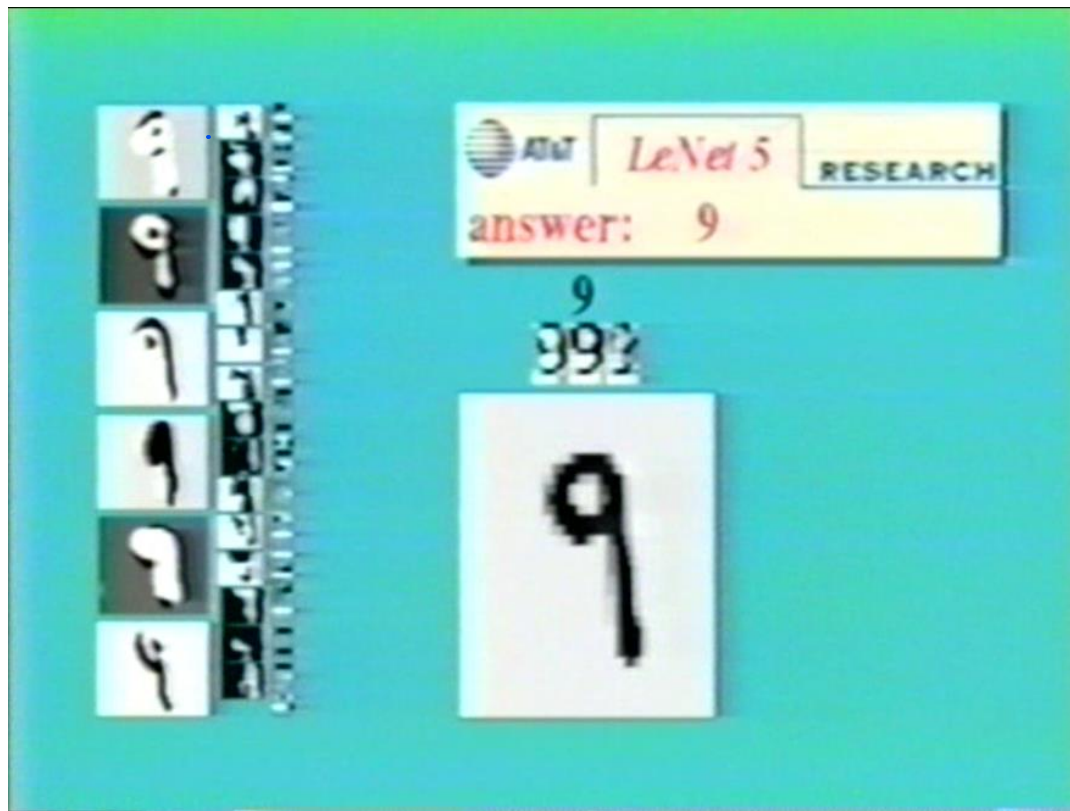
$$h_{\Theta}(x) \in \mathbb{R}^4$$

Want $h_{\Theta}(x) \approx \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$, $h_{\Theta}(x) \approx \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$, $h_{\Theta}(x) \approx \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$, etc.
when pedestrian when car when motorcycle

Training set: $(x^{(1)}, y^{(1)}), (x^{(2)}, y^{(2)}), \dots, (x^{(m)}, y^{(m)})$

$y^{(i)}$ one of $\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$
pedestrian car motorcycle truck

Handwritten digit classification with Convolutional NN (CNN)



Handwritten digit classification with ConvNet

