

Neural Networks - Deep Learning

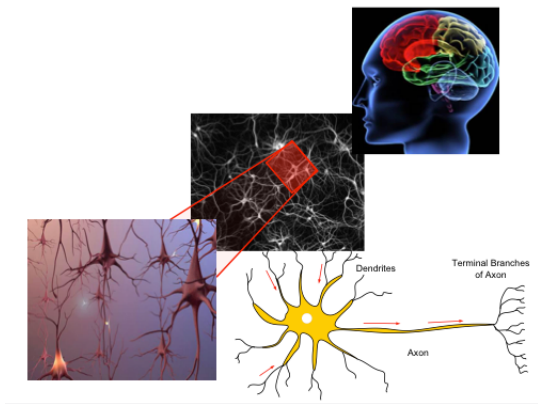
Artificial Intelligence @ Allegheny College

Janyl Jumadinova

March 20-23, 2023

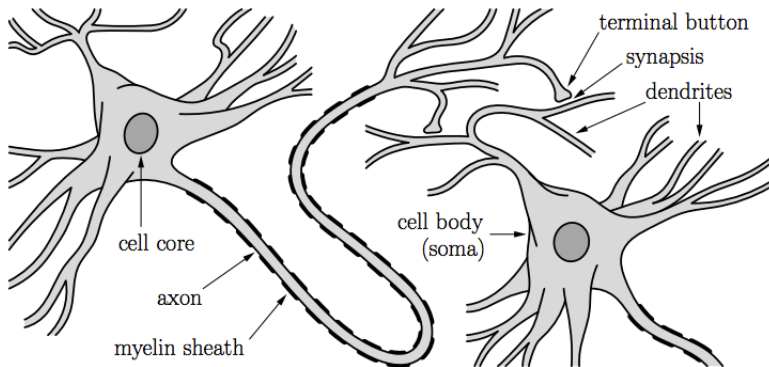
Credit: Google Workshop

Neural Networks



Neural Networks

Structure of a prototypical biological neuron



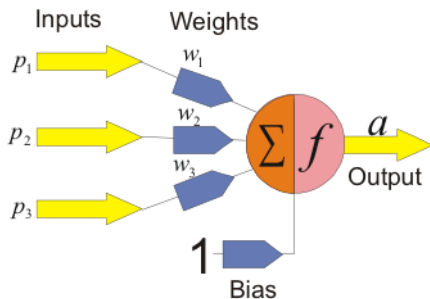
Neural Networks

Neural computing requires a number of **neurons**, to be connected together into a **neural network**.

Neurons are arranged in layers.

Two main **hyperparameters** that control the architecture or topology of the network: 1) the number of layers, and 2) the number of nodes in each hidden layer.

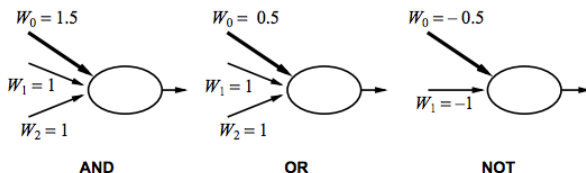
Neural Networks



$$a = f(p_1w_1 + p_2w_2 + p_3w_3 + b) = f(\sum p_iw_i + b)$$

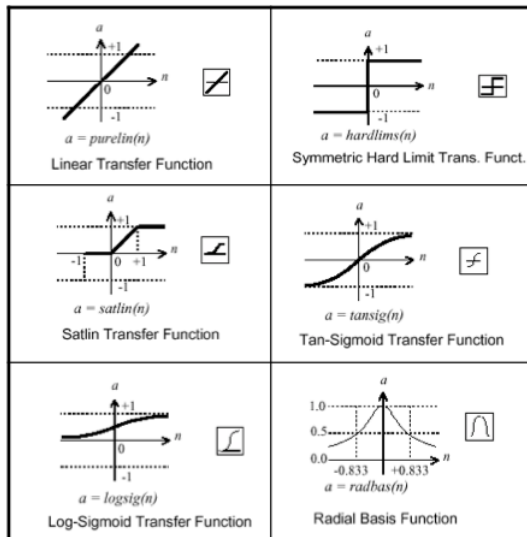
Activation Functions

- The activation function is generally non-linear.
- Linear functions are limited because the output is simply proportional to the input.



McCulloch and Pitts: every Boolean function can be implemented

Activation Functions

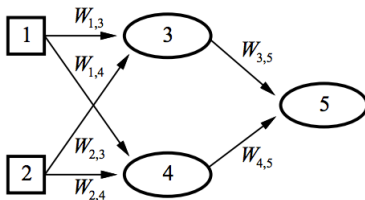


Network structures

Two phases in each iteration:

- ① Calculating the predicted output \mathbf{y} , known as feed-forward
- ② Updating the weights and biases, known as backpropagation

Feed-forward example



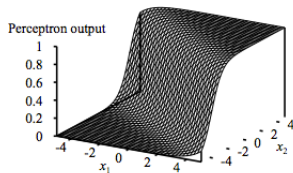
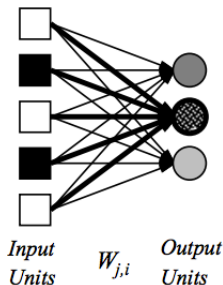
Feed-forward network = a parameterized family of nonlinear functions:

$$\begin{aligned} a_5 &= g(W_{3,5} \cdot a_3 + W_{4,5} \cdot a_4) \\ &= g(W_{3,5} \cdot g(W_{1,3} \cdot a_1 + W_{2,3} \cdot a_2) + W_{4,5} \cdot g(W_{1,4} \cdot a_1 + W_{2,4} \cdot a_2)) \end{aligned}$$

Feed-forward networks:

- Single-layer perceptrons
- Multi-layer perceptrons

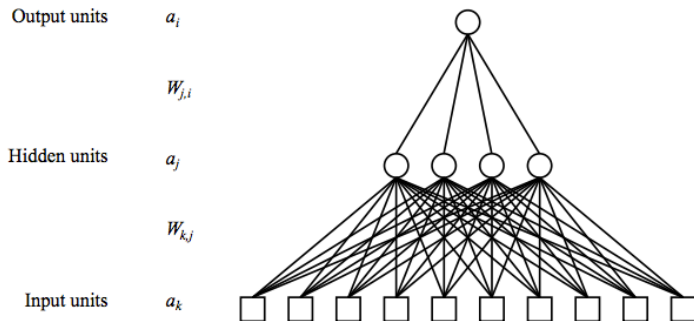
Single-layer Perceptrons



Output units all operate separately – no shared weights.

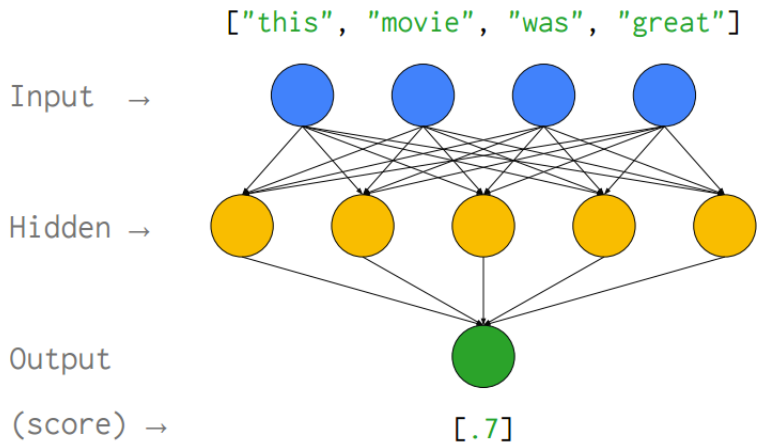
Adjusting weights moves the location, orientation, and steepness of cliff.

Multi-layer Perceptrons

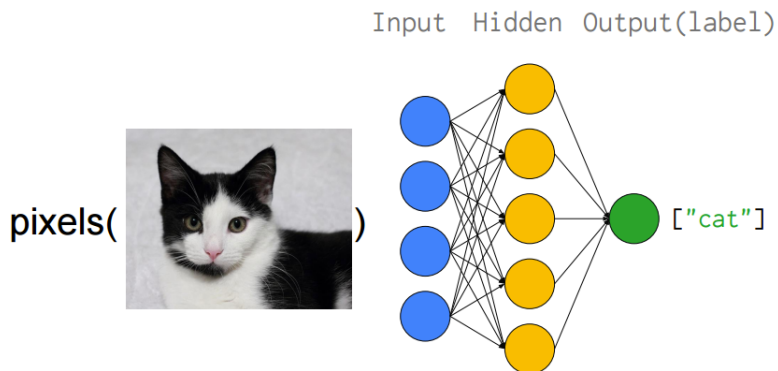


- Layers are usually fully connected.
- Numbers of hidden units typically chosen by hand.

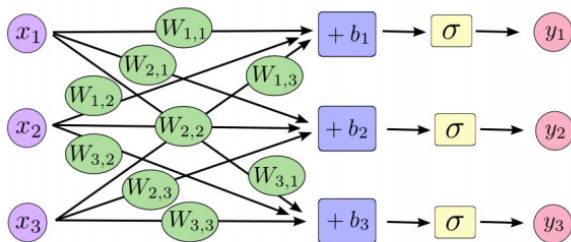
Neural Networks



Neural Networks



Neural Networks: A fully connected NN layer



$$\begin{aligned} y_1 &= \sigma (W_{1,1}x_1 + W_{1,2}x_2 + W_{1,3}x_3 + b_1) \\ y_2 &= \sigma (W_{2,1}x_1 + W_{2,2}x_2 + W_{2,3}x_3 + b_2) \\ y_3 &= \sigma (W_{3,1}x_1 + W_{3,2}x_2 + W_{3,3}x_3 + b_3) \end{aligned}$$

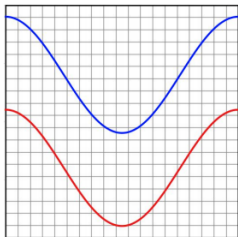
Implementation as Matrix Multiplication

$$\begin{aligned} y_1 &= \sigma (W_{1,1}x_1 + W_{1,2}x_2 + W_{1,3}x_3 + b_1) \\ y_2 &= \sigma (W_{2,1}x_1 + W_{2,2}x_2 + W_{2,3}x_3 + b_2) \\ y_3 &= \sigma (W_{3,1}x_1 + W_{3,2}x_2 + W_{3,3}x_3 + b_3) \end{aligned}$$

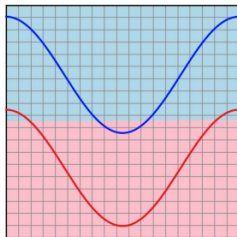
$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \sigma \left(\begin{bmatrix} W_{1,1} & W_{1,2} & W_{1,3} \\ W_{2,1} & W_{2,2} & W_{2,3} \\ W_{3,1} & W_{3,2} & W_{3,3} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right)$$

Non-Linear Data Distributions

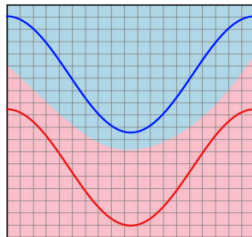
The Problem



Linear Model



Neural Network



Deep Learning

- Most current machine learning works well because of human-designed representations and input features.
- Machine learning becomes just optimizing weights to best make a final prediction.

Deep Learning

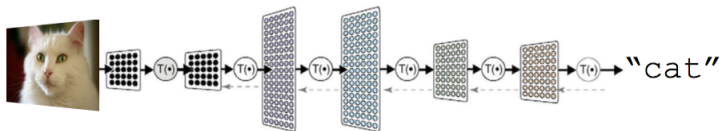
- Most current machine learning works well because of human-designed representations and input features.
- Machine learning becomes just optimizing weights to best make a final prediction.
- **Deep learning** algorithms attempt to learn multiple levels of representation of increasing complexity/abstraction.

Deep Learning

- Each neuron implements a relatively simple mathematical function.
- $y = g(\overline{w} \cdot \overline{x} + b)$

Deep Learning

- Each neuron implements a relatively simple mathematical function.
- $y = g(\overline{w} \cdot \overline{x} + b)$
- The composition of $10^6 - 10^9$ such functions is powerful.



Deep Learning

Book: <http://www.deeplearningbook.org/>

Chapter 5

“A core idea in deep learning is that we assume that the data was generated by the composition of factors or features, potentially at multiple levels in a hierarchy.”

Deep Learning

Results get better (to a degree) with:

- more data
- bigger models
- more computation

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Better algorithms, new insights and improved methods help, too!

TensorFlow



- Open source Machine Learning library
- Especially useful for **Deep Learning**
- For research **and** production
- **Apache 2.0** license
- [tensorflow.org](https://www.tensorflow.org)

- *Epoch*: a training iteration (one pass through the dataset).
- *Batch*: Portion of the dataset (number of samples after dataset has been divided).
- *Regularization*: a set of techniques that helps learning models to converge
(<http://www.godeep.ml/regularization-using-tensorflow/>).

TensorFlow

A multidimensional array.



- Operates over **tensors**: n-dimensional arrays A graph of operations.

TensorFlow

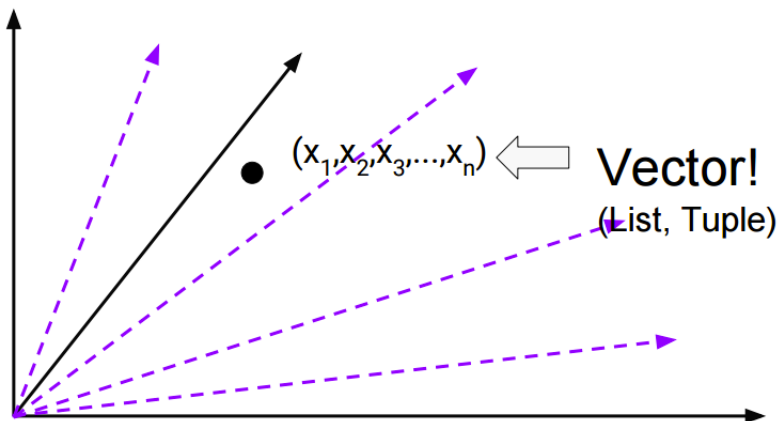
A multidimensional array.



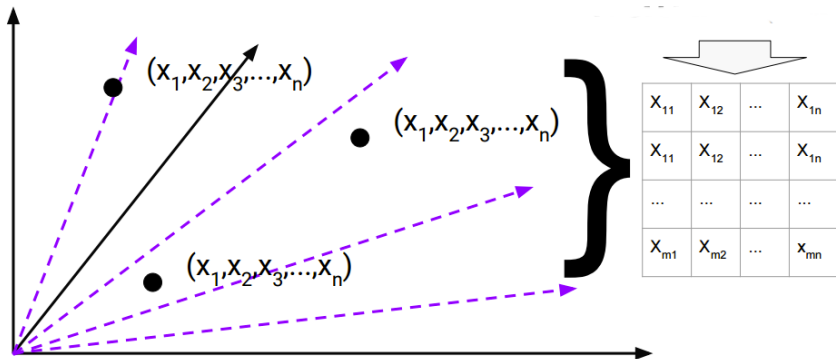
- Operates over **tensors**: n-dimensional arrays
- Using a **flow graph**: data flow computation framework

- 5.7 ← Scalar
- Number, Float, etc.

TensorFlow



TensorFlow

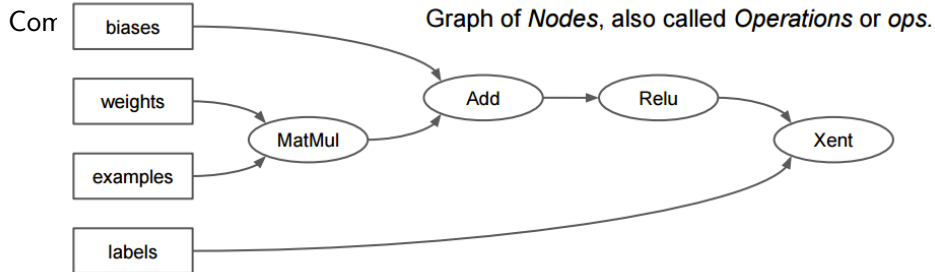


- Tensors have a **Shape** that is described with a vector

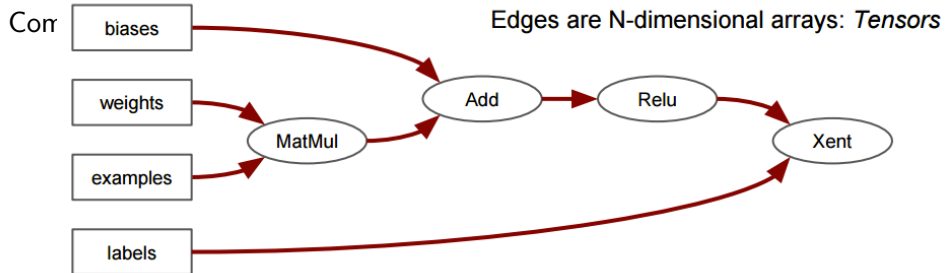
TensorFlow

- Tensors have a **Shape** that is described with a vector
- [1000, 256, 256, 3]
- 10000 Images
- Each Image has 256 Rows
- Each Row has 256 Pixels
- Each Pixel has 3 values (RGB)

TensorFlow



TensorFlow



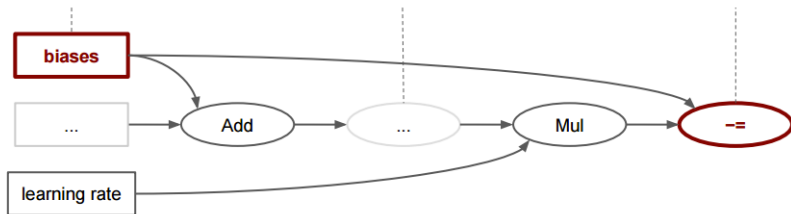
TensorFlow

Computation is a dataflow graph with state

'Biases' is a variable

Some ops compute gradients

--= updates biases



Core TensorFlow data structures and concepts

- **Graph:** A TensorFlow computation, represented as a dataflow graph:
 - collection of ops that may be executed together as a group.

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- **Graph:** A TensorFlow computation, represented as a dataflow graph:
 - collection of ops that may be executed together as a group.
- **Operation:** a graph node that performs computation on tensors
- **Tensor:** a handle to one of the outputs of an Operation:
 - provides a means of computing the value in a TensorFlow Session.

- **Constants**

TensorFlow

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- **Placeholders**: must be fed with data on execution.
- **Variables**: a modifiable tensor that lives in TensorFlow's graph of interacting operations.
- **Session**: encapsulates the environment in which Operation objects are executed, and Tensor objects are evaluated.

TensorFlow

<i>Category</i>	<i>Examples</i>
Element-wise math ops	Add , Sub, Mul , Div, Exp, Log, Greater, Less...
Matrix ops	Concat , Slice, Split , Constant, Rank, Shape ...
Matrix ops	MatMul , MatrixInverse, MatrixDeterminant...
Stateful ops	Variable , Assign, AssignAdd...
NN building blocks	SoftMax , Sigmoid, ReLU , Convolution2D ...
Checkpointing ops	Save, Restore
Queue & synch ops	Enqueue, Dequeue, MutexAcquire...
Control flow ops	Merge, Switch, Enter, Leave...

`https://playground.tensorflow.org`