

Basics of Artificial Neural Networks (ANN)

Intro

- Intro
- Properties
- Application

Intro

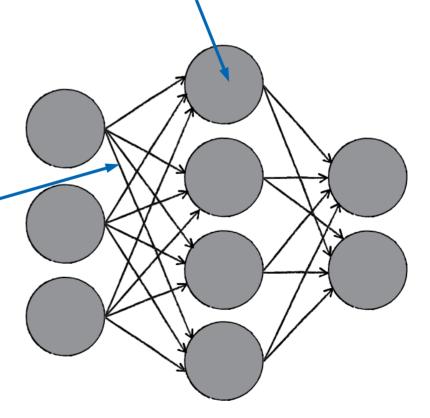
- Intro
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Intro/Definition

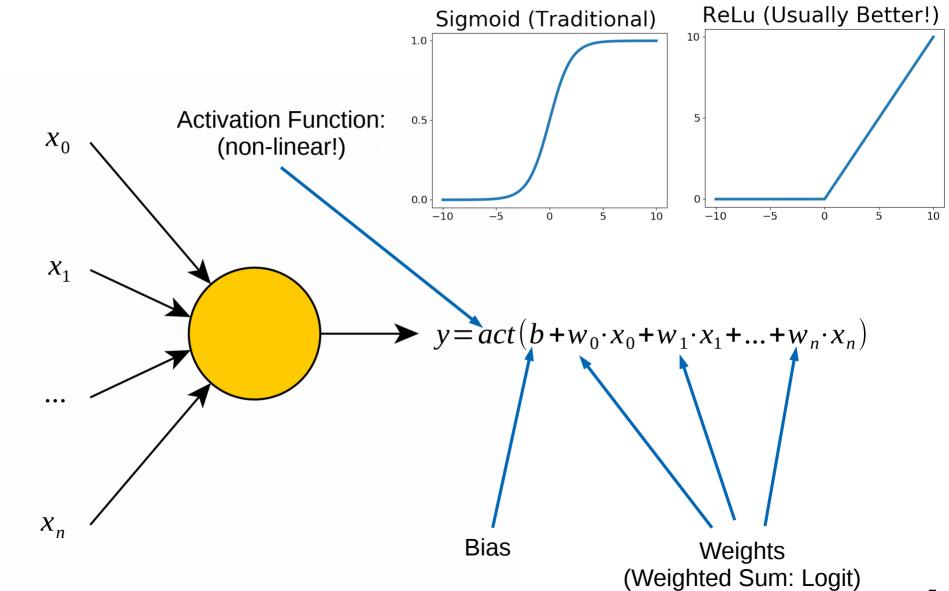
ANN:

Network of small Computation Units (Neurons)

- Inspired by Brain
- Computation Model
- Not a Simulation
- Connections: Synapses



Intro/Single Perceptron



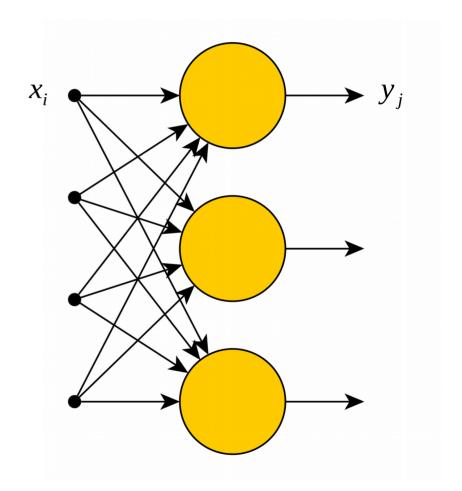
Intro/Layers

- 1 Output/Perceptron
- → Use multiple
- Organized in Layers
- Weighted sum

$$y_{j} = act \left(b_{i} + \sum_{i} w_{ij} \cdot x_{i} \right)$$

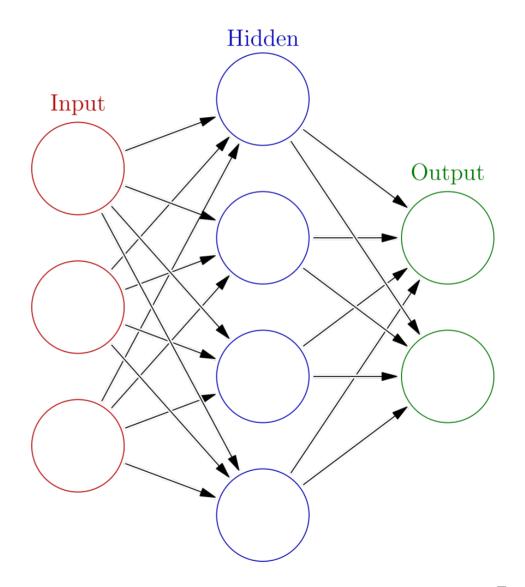
→ Matrix multiplication:

$$\vec{y} = act(\vec{b} + W^T \cdot \vec{x})$$



Intro/Multi-Layer Perceptron

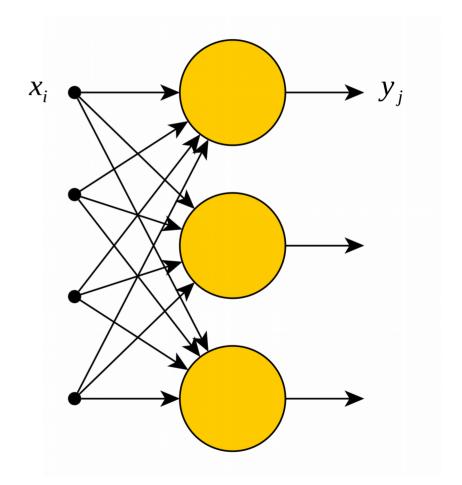
- Layers Stackable
- → Deep NN



Intro/Softmax Layer

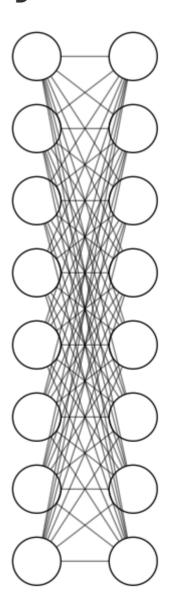
$$y_j = \frac{e^{b_j + w_j \cdot x_j}}{\sum_{i} e^{b_i + w_i \cdot x_i}}$$

- Sums up to 1
- → Probability-Like
- → Classification



Intro/Convolutional Layer

- Fully Connected Layer
- 10³ Neurons Each
- → 10⁶ Synapses
- → Reduce Synapes to "Neighborhood"
- Works well for Images or Sequences

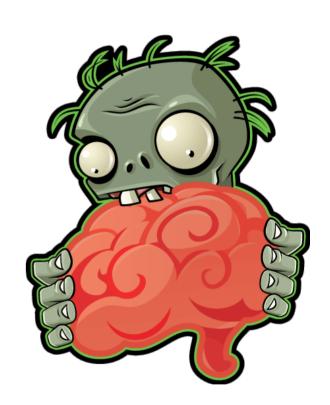


Properties

- Intro
- Properties
- Application

Properties

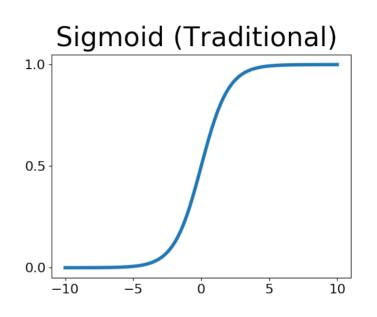
- Why are ANN so powerful/popular?
- They're like a human brain?
 - Not really... but...
- Model any (mathematical) Function!
- Some ANN Touring complete (model any computer program)
- They are trainable (/fittable/optimizable/programmable)
- Think of a computer program that writes itself
 - albeit not perfectly...



Properties/Boolean Logic

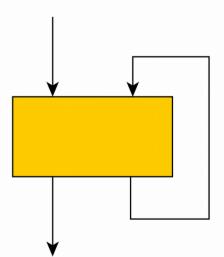
$$a \lor b = sig(10(a+b)-5)$$

 $sig(10(0+0)-5) \approx 0 \checkmark$
 $sig(10(0+1)-5) \approx 1 \checkmark$
 $sig(10(1+0)-5) \approx 1 \checkmark$
 $sig(10(1+1)-5) \approx 1 \checkmark$
 $\neg a = sig(5-10 \cdot a)$
 $sig(5-10 \cdot 0) \approx 1 \checkmark$
 $sig(5-10 \cdot 1) \approx 0 \checkmark$
 $a \land b = \neg(\neg a \lor \neg b)$



Properties/Memory

- We've got boolean logic!
- Whats missing for Touring Completeness?
 - Memory!
- → Recurrent Neural Networks RNN
- Some outputs fed back as inputs
- Special (forgetful) cells
 - -LSTM (Long Short-Term Memory)
 - -GRU, ...



Properties/Trainability

- Outputs of ANN can be trained
- Requires loss/error function, e.g. mean squared error
- ANN well derivable → fast optimization
 - Computed automaticall by Tensorflow, PyTorch, ...
- (Stochastic) Gradient Descent (SGD) as Opimizer
 - better than complicated Methods
 - Process: Feed-Forward & Back-Propagation

$$\vec{y} = \vec{x}^{L_n}$$

$$\vec{x}^{L_i} = act \left(\vec{b}^{L_i} + (W^{L_i})^T \cdot \vec{x}^{L_{i-1}} \right)$$

$$\frac{\partial \vec{x}^{L_i}}{\partial W^{L_j}} = act' \left(\vec{b}^{L_i} + (W^{L_i})^T \cdot \vec{x}^{L_{i-1}} \right) \cdot (W^{L_i})^T \cdot \frac{\vec{x}^{L_{i-1}}}{\partial W^{L_j}}$$



That's it! - Questions?

Sources

- Definition ANN [Slide 4]: http://natureofcode.com/book/chapter-10-neural-networks/
- MLP Layers [Slide 7]: https://en.wikipedia.org/wiki/Artificial neural network
- Zombie Brain [Slide 11]: http://plantsvszombies.wikia.com/wiki/File:HDZombieAndBrain.png
- Remaining images created with yEd https://www.yworks.com/products/yed