

# Al Society Meeting Spring 2020

Basics of Deep Learning

#### Curriculum

- Week 7 -> Basics of Deep Learning
- Week 8 -> Basics of Natural Language Processing
- Week 9 -> Long Short Term Memory/Recurrent Neural Networks

#### **Outline**

- 1. Artificial Neural Networks (ANNs) and Deep Learning
- 2. Layers
- 3. Important Terms
- 4. Different Neural Network Architectures
- 5. Applications
- 6. Exercises

## Artificial Neural Networks (ANNs) and Deep Learning

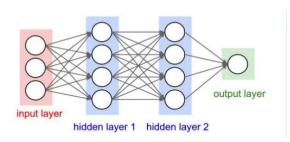
#### **ANNs**

- ANNs have units of calculation called **neurons** (connected by weighted values).
- What is a neuron? Consists of an activation function + weighted connections
- Long story short:
  - neuron will perform some sort of a calculation, possibly through an activation function, and then the result of this calculation will be multiplied by a weight as it passes through.
  - The weighted result can sometimes be the output of your neural network or you can have more neurons configured in layers,
     which is the basic concept to deep learning.
- When we ask the system a large number of questions that's how the neurons learn how to carry out any specific calculations (recall supervised learning) and provide them with model/expected answers.
- Keep in mind, Deep Learning always has more than 1 hidden layer (you can jot down the dots once you reach the slides on Layers).
- Deep learning is a branch of Machine Learning which uses different types of neural network architectures. These algorithms mimic the way our brain functions.

#### Layers

#### Layers

- Layers are a set of neurons which help us determine the number of neurons we can use.
- The 3 different kinds:
  - Input: Data provided to the neural network
  - Hidden: Depends on your architecture and processes **activation functions**
  - Output: Provide us with the results/finished computations
- This hierarchy helps the output layer with predictions.
- Each layer extracts higher features from the input layer until the final layer makes a decision about what the input shows.
- More layers = higher level features can be learned



#### **Important Terms**

#### Important terms

- <u>Linear Algebra behind DL</u>
- Activation function
  - Neurons detect features and transform them through these non-linear activation functions before passing them onto the output layer; needed to learn complex/non-linear representations of data. Example: Sigmoid function.
- Neural networks learn weights and biases using the gradient descent algorithm -- how are the gradients of the
  cost functions computed or really, how does the neural network learn? Backpropagation! More on math behind
  backpropagation <a href="here">here</a>.

#### Important terms (continued)

- More on backpropagation:
  - Used to find the weights/biases that lead to the most accurate output for the NN, where the weights are initialized randomly at most times.
  - These weights and biases keep adjusting depending on what the model learns.
  - Gradient descent is then used to come up with a minimized cost function (talked about in our previous sessions) using these weights.
  - The "training" of the model comprises of working with more and more data to give you a neural network with a high accuracy rate.
    - Output from the neural network -> calculate cost -> backpropagate to tune weights
  - The most important ingredient? Data. If you don't have clean data, your model (no matter how optimal it is) can only help you so much.

### Different Neural Network Architectures

#### Different Neural Network Architectures

- Recurrent Neural Networks
  - Help in understanding context based on previous behavior (used in Natural Language Processing) and in retaining information (example: LSTM) by going back through the process and building on top of it
- Convolutional Neural Networks
  - Help in reducing the number of parameters that need to be optimized
  - Notes symmetry in how the neurons are connected; helps in recognizing patterns
  - Commonly used in working with images by recognizing patterns around pixels
  - More on CNNs
- Feel free to look more into these
  - Feedforward Neural Network
  - Radial Basis Function Neural Network
  - Multilayer perceptron
  - Modular Neural Networks

#### **Applications**

- Speech Recognition (NLP)
- Image Recognition (computer vision, try ECS 174 to learn more)
- Ads, web search
- Notable people who have contributed significantly: Geoffrey Hinton, Andrew Ng, Yann LeCun, Ian Goodfellow (one of the authors of the deep learning book), Jeff Dean

#### **Exercises**

- Tinker with a Neural Network <u>here</u>
- Try this Neural Network exercise <u>here</u>
- Try this Sentiment Analysis exercise by following this article

# Thank you! Next week: Basics of Natural Language Processing