



# Machine Learning Crash Course



Week 5





I don't think that artificial intelligence means doomsday, and I think many new jobs will be created, too. However, it is becoming increasingly unlikely that these new types of jobs will favor low-income demographics. We need to address the needs of those who will be left out of the new job market.” ~ Jens Martin Skibsted



Nothing is more important, certainly during these times of artificial intelligence, than our public education. And as it continues to grow and evolve, I think you and I know this is going to be critical that we are constantly training and retraining and creating these next-generation jobs.” ~ Marc Benioff



# Topics

- ◇ Neural Networks
  - Cost Function
  - Overview
  - Backpropagation Algorithm
  - Intuition
- ◇ Implementation
  - Unrolling Parameters
  - Gradient Checking
  - Initializations
- ◇ Summary



A decorative graphic on the left side of the slide. It features a large, light blue hexagon in the center, surrounded by several smaller hexagons in various shades of blue and cyan. These smaller hexagons contain white icons: a lightbulb, a thumbs-up, a smartphone, a magnifying glass, a gear, and a speech bubble. There is also a small network diagram icon with a central node and five connecting lines.

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# Neural Networks



$$J(\theta) = -\frac{1}{m}(\sum_{i=1}^m y^i \log h_{\theta}(x^i) + (1 - y^i) \log(1 - h_{\theta}(x^i))) + \frac{\lambda}{2m} \sum_{j=1}^n \theta_j^2$$

# Cost Function

- ◇ Neural Network cost can be conceptualized in 2 parts:
  - For  $\forall$  “training” data (1 -> m):
    - Sum(each element of the “output” vector)
  - Calculate the weight decay term
- ◇ DEMO





# Overview

- ◇ Forward Propagation
  - Pushes the input through the NN
- ◇ Back Propagation
  - Receives the NN output
  - Receives the  $H(\theta) - y$  from a given layer  $L$
  - Partial Derivatives & gradient descent to minimise the cost function  $J(\theta)$





# Backpropagation

- ◇ Goal to find parameters  $\theta$  that  $\min(J(\theta))$ 
  - Calculate the **error** of the nodes in each layer
  - Calculate the **activation** of the nodes in each layer
  - Work backwards from the output layer
  - Obtain the **deltas** to calculate the partial derivative (**gradient**) of the cost function
- ◇ (DEMO)





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# Implementation



# Unrolling Parameters

- ◇ Advanced Optimization Technique
  - Unroll the matrices into vectors for the algorithm that will  $\min(\text{cost})$
- ◇ DEMO





# Gradient Checking

- ◇ Verification that the cost function is decreasing over time
- ◇ Additional methodology to check for correct implementation
- ◇ Disable when training a given model due to time complexity
- ◇ **DEMO**





# Initializations

- ◇ Initializing  $\theta$ 
  - Random Initializations
- ◇ Initializing  $\alpha$ 
  - Using an Advanced Optimization Algorithm
- ◇ **DEMO**



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# Summary



# Summary

- ◇ Init weights randomly
- ◇ Implement forward propagation
- ◇ Implement cost function
- ◇ Implement backpropagation
- ◇ Utilize gradient checking
- ◇ Utilize advanced optimization
- ◇ **“CODE” DEMO**
- ◇ **<https://github.com/TensorFlow-ML-Architectures/TensorFlow-Multilayer-Perceptron>**





# Thanks!

## Any questions?

You can find me at:

<http://cs.oswego.edu/~kzeller>





# Sources

- ◇ <https://developers.google.com/machine-learning/crash-course/introduction-to-neural-networks/video-lecture>
- ◇ [https://www.cs.toronto.edu/~tijmen/csc321/slides/lecture\\_slides\\_lec3.pdf](https://www.cs.toronto.edu/~tijmen/csc321/slides/lecture_slides_lec3.pdf)
- ◇ <https://www.ritchieng.com/>
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- ◇ <https://mattmazur.com/2015/03/17/a-step-by-step-backpropagation-example/>

