

Info 251: Applied Machine Learning  
Lab 10  
4/1/2020

# Topics

- ▶ Random Forests
- ▶ Neural networks
- ▶ Tensorflow

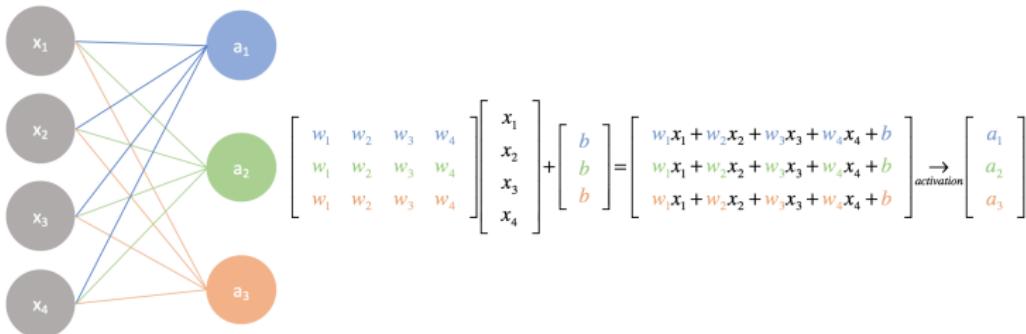
# Neural Networks

- ▶ est 1940s
- ▶ Great success in AI
- ▶ Image recognition, speech processing etc

# Neural Networks

Input layer      Output layer

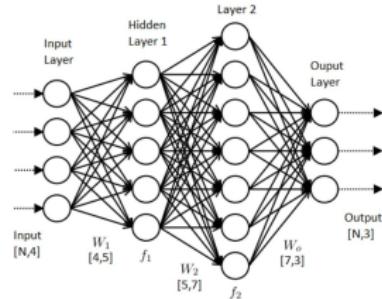
A simple neural network



- ▶ layer 1 → input:  $x$ , output  
 $a = g(W_1x + b_1)$ ,  $W \in \mathbb{R}^{3 \times 4}$ ,  $b \in \mathbb{R}^3$
- ▶ layer 2 → input:  $g(W_1x + b_1)$ , output  $g(W_2g(W_1x + b_1) + b_2)$
- ▶ ...

# Neural Networks

- ▶ Universal function approximation theorem
- ▶ However, we don't know how many layers or their size
- ▶ How choose size then?
- ▶ Input layer
- ▶ Output layer
- ▶ Hidden layers
  - If data linearly separable need none!
  - Usually one is enough
  - Extra layers usually improve but add computational cost
  - Size of hidden layers
  - ROT: in between the size of input and output.
- ▶ Start with the above and keep iterating
- ▶ If we have unlimited computational power why not increase the dimensions a lot?

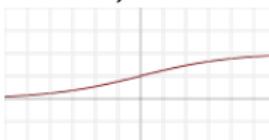


# Neural Networks

- ▶ How about activation functions?
- ▶ "Identity"  $g(x) = x$



- ▶ "sigmoid"  $g(x) = 1/(1 + e^{-x})$  if  $x > 0$  and 0 o.w.



- ▶ "Tanh"  $g(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$



- ▶ "ReLU"  $g(x) = x$  if  $x > 0$  and 0 o.w.



# Neural Networks

- ▶ Choice of activation function?

# Neural Networks

- ▶ Choice of activation function? **Trial and error**

## Neural Networks

- ▶ Choice of activation function? **Trial and error**
- ▶ Objective function?
- ▶ Commonly, Regression → MSE  $\frac{1}{n} \sum_{i=1}^n (y_i - \tilde{y}_i)^2$ ,  
Classification → CE  $\sum_{i=1}^n -y_i \log \tilde{y}_i - (1 - y_i) \log(1 - \tilde{y}_i)$

# Neural Networks

- ▶ Choice of activation function? **Trial and error**
- ▶ Objective function?
- ▶ Commonly, Regression → MSE  $\frac{1}{n} \sum_{i=1}^n (y_i - \tilde{y}_i)^2$ ,  
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- ▶ Data pre-processing
- ▶ Normalization?

# Neural Networks

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- ▶ Data pre-processing
- ▶ Normalization? Yes! Better for optimization
- ▶ Categorical variables?
- ▶ Regularization?

# Neural Networks

- ▶ How do we optimize
- ▶ GD?
- ▶ Tough composition of functions
- ▶ Backpropagation

## NN: Implementation

- ▶ Tensorflow
- ▶ Python and C++ under the hood
- ▶ Create dataflow graphs-structures that describe how data moves through a graph
- ▶ Nodes represent mathematical operations
- ▶ Connections or edges between nodes are a multidimensional data arrays, or tensors
- ▶ Can use GPUs for computations
- ▶ (Remove) the feed dictionary specifies the placeholder values for that computation
- ▶ Other choices PyTorch, Caffe etc

- ▶ Notebook