

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

Topics

- ▶ Random Forests
- ▶ Neural networks
- ▶ Tensorflow

Neural Networks

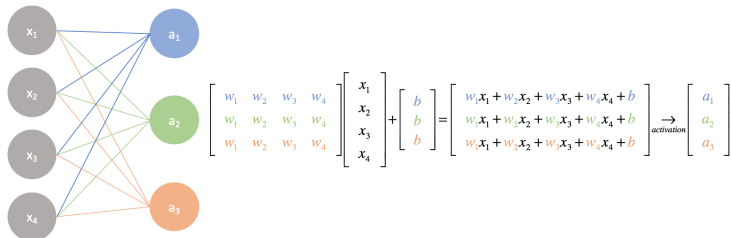
- ▶ est 1940s
- ▶ Great success in AI
- ▶ Image recognition, speech processing etc

Neural Networks

Input layer

Output layer

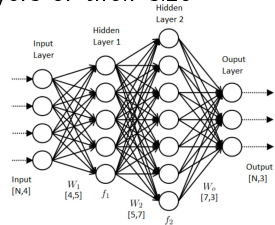
A simple neural network



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



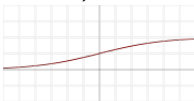
Neural Networks

- ▶ How about activation functions?

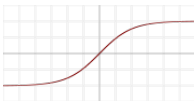
- ▶ "Identity" $g(x) = x$



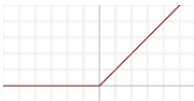
- ▶ "sigmoid" $g(x) = 1/(1 + e^{-x})$ if $x > 0$ and 0 o.w.



- ▶ "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 $\rightarrow \text{MSE } \frac{1}{n} \sum_{i=1}^n (y_i - \tilde{y}_i)^2$,
Classification $\rightarrow \text{CE } \sum_{i=1}^n -y_i \log \tilde{y}_i - (1 - y_i) \log (1 - \tilde{y}_i)$

Neural Networks

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

Neural Networks

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