Standard notations for Deep Learning

This document has the purpose of discussing a new standard for deep learning mathematical notations.

1 Neural Networks Notations.

General comments:

 $\cdot\,$ superscript (i) will denote the i^{th} training example while superscript [l] will denote the l^{th} layer

Sizes:

 $\cdot m$: number of examples in the dataset

 $\cdot n_x$: input size

 $\cdot n_y$: output size (or number of classes)

 $\cdot n_h^{[l]}$: number of hidden units of the l^{th} layer

In a for loop, it is possible to denote $n_x = n_h^{[0]}$ and $n_y = n_h^{[number of layers +1]}$.

 $\cdot L$: number of layers in the network.

Objects:

 $\cdot X \in \mathbb{R}^{n_x \times m}$ is the input matrix

 $\cdot x^{(i)} \in \mathbb{R}^{n_x}$ is the i^{th} example represented as a column vector

 $\cdot Y \in \mathbb{R}^{n_y \times m}$ is the label matrix

 $y^{(i)} \in \mathbb{R}^{n_y}$ is the output label for the i^{th} example

- $W^{[l]} \in \mathbb{R}^{\text{number of units in next layer } \times \text{ number of units in the previous layer}}$ is the weight matrix, superscript [l] indicates the layer
- $b^{[l]} \in \mathbb{R}^{\text{number of units in next layer}}$ is the bias vector in the l^{th} layer
- $\hat{y} \in \mathbb{R}^{n_y}$ is the predicted output vector. It can also be denoted $a^{[L]}$ where L is the number of layers in the network.

Common forward propagation equation examples:

 $a=g^{[l]}(W_x x^{(i)}+b_1)=g^{[l]}(z_1)$ where $g^{[l]}$ denotes the l^{th} layer activation function

 $\hat{y}^{(i)} = softmax(W_h h + b_2)$

- · General Activation Formula: $a_j^{[l]} = g^{[l]}(\sum_k w_{jk}^{[l]} a_k^{[l-1]} + b_j^{[l]}) = g^{[l]}(z_j^{[l]})$
- · J(x, W, b, y) or $J(\hat{y}, y)$ denote the cost function.

Examples of cost function:

$$J_{CE}(\hat{y}, y) = -\sum_{i=0}^{m} y^{(i)} \log \hat{y}^{(i)}$$

$$J_1(\hat{y}, y) = \sum_{i=0}^{m} |y^{(i)} - \hat{y}^{(i)}|$$

$\mathbf{2}$ **Deep Learning representations**

For representations:

- $\cdot\,$ nodes represent inputs, activations or outputs
- · edges represent weights or biases

Here are several examples of Standard deep learning representations



 $x^{(i)}$ Reli $x_{1}^{(i)}$ $x_2^{(i)}$ $(a_{2}^{l_{1}})$ ŷ $(a_1^{[2]})$ x⁽ⁱ⁾ 12287 $x_{12288}^{(i)}$

Figure 1: Comprehensive Network: representation commonly used for Neural Networks. For better aesthetic, we omitted the details on the parameters $(w_{ij}^{[l]}$ and $b_i^{[l]}$ etc...) that should appear on the edges