

High Capacity Neural Network

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1 Introduction

TODO(domenic): Write this section

2 Related Literature

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2.1 Forward Computation

The forward computation of a fully connected feedforward neural network with N layers can be described as follows:

$$x_{n-1} = a_n(t_n(x_n)) \quad (1)$$

$\{N \in \mathbb{N} | N > 0\}$

$n \in N$

x_n is a vector

x_N is the input vector

$\hat{y} = x_1$ is the output vector

$t_n(x)$ is a transform function

$a_n(x)$ is an activation function

The standard transform function, $t_n(x)$, is:

$$t_n(x) = f_n(x) \quad (2)$$

$$f_n(x) = x \cdot W_n + b_n \quad (3)$$

$W_n \in \mathbb{R}^2$

$b_n \in \mathbb{R}$

The transform function proposed in this paper is:

$$t_n(x) = f_n(x) \cdot g(x) \quad (4)$$

$$g_n(x) = x \cdot Y_n + c_n \quad (5)$$

$Y_n \in \mathbb{R}^2$

$c_n \in \mathbb{R}$

Rather than using the standard transform, (2), a quadratic transform, (4), is performed. The quadratic transform is a superset of the standard design, which becomes apparent when $Y_n \in \{0, \dots, 0\}$ and $c_n \in \{1, \dots, 1\}$. Using a quadratic transformation enables the neural network to represent any bounded degree polynomial over an infinite domain using a finite number of nodes. The standard design would require an infinite number of nodes to do the same.

2.2 Backward Computation

The back propagation algorithm is slightly different from that of network using the standard transform. However, it is mostly the same. The total network error is calculated as follows:

$$E = \sum_i c(y_i, \hat{y}_i) \quad (6)$$

Where $c(y, \hat{y})$ is a cost function. For example, a squared error cost function would look like:

$$c(y, \hat{y}) = \frac{(y - \hat{y})^2}{2} \quad (7)$$

$$c'(y, \hat{y}) = y - \hat{y} \quad (8)$$

The key thing about back propagation is that its a dynamic program and caches the part of the gradient calculation that will be used upstream.

C_n is the matrix of cached calculations at level n

$$C_1 = c'(y, \hat{y})$$

$$C_{n+1} = a'_n(x_n) \cdot (C_n \cdot t'_n(x_n)^\top) \quad (9)$$

3 Experiments

TODO(domenic): Write this section

4 Conclusion

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