

If you wish to make a neural network from scratch...

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Neural network

Fancy non-linear least squares fitting

Minimize $\chi^2 = \frac{1}{N} \sum_{j=0}^{N-1} [y_j - p(x_j)]^2 \equiv \frac{1}{N} \sum_{j=0}^{N-1} s_j$

For $p(x) = \phi \left(\sum_{k=0}^{M-1} a_k \phi(b_k x + c_k) + d \right)$

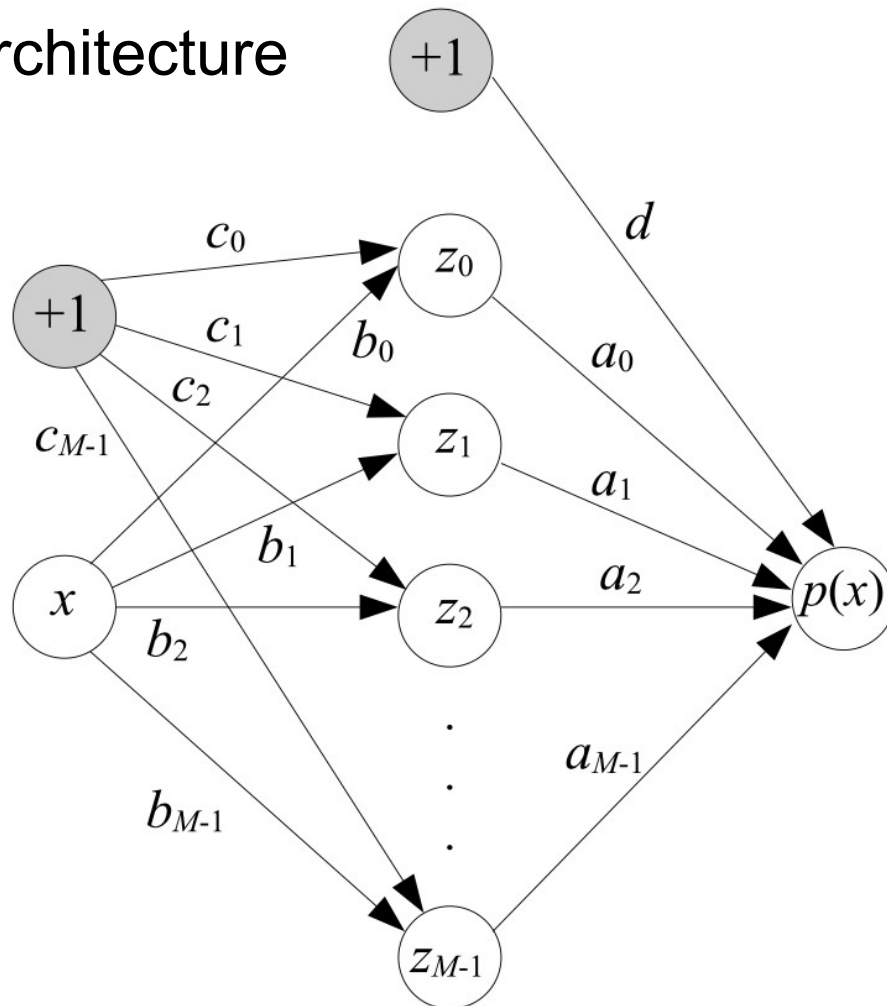
If it helps, consider

$$p(x) = \phi \left(\sum_{k=0}^{M-1} a_k z_k + d \right), \quad z_k = \phi(b_k x + c_k)$$

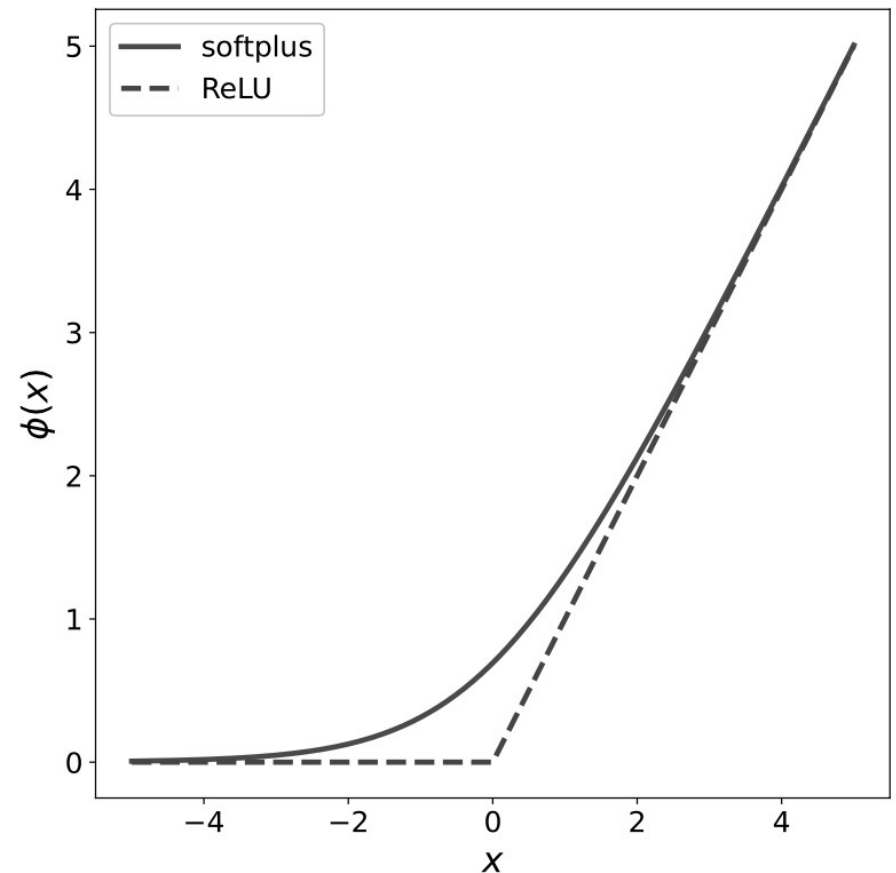
Neural network

$$p(x) = \phi \left(\sum_{k=0}^{M-1} a_k z_k + d \right), \quad z_k = \phi(b_k x + c_k)$$

Architecture



Activation function $\phi(x) = \ln(1 + e^x)$



Neural network

Determine your parameters via

Gradient descent: $\mathbf{w}^{(i)} = \mathbf{w}^{(i-1)} - \gamma \nabla \chi^2(\mathbf{w}^{(i-1)}) = \mathbf{w}^{(i-1)} - \frac{\gamma}{N} \sum_{j=0}^{N-1} \nabla s_j(\mathbf{w}^{(i-1)})$

Stochastic gradient descent: $\mathbf{w}^{(i)} = \mathbf{w}^{(i-1)} - \gamma \nabla s_j(\mathbf{w}^{(i-1)})$

More explicitly:

$$\begin{aligned} a_m^{(i)} &= a_m^{(i-1)} - \gamma \left. \frac{\partial s_j}{\partial a_m} \right|^{(i-1)}, & b_m^{(i)} &= b_m^{(i-1)} - \gamma \left. \frac{\partial s_j}{\partial b_m} \right|^{(i-1)}, \\ c_m^{(i)} &= c_m^{(i-1)} - \gamma \left. \frac{\partial s_j}{\partial c_m} \right|^{(i-1)}, & d^{(i)} &= d^{(i-1)} - \gamma \left. \frac{\partial s_j}{\partial d} \right|^{(i-1)} \end{aligned}$$

Neural network

Backpropagation: $\frac{\partial s_j}{\partial a_m} = 2[p(x_j) - y_j] \frac{\partial p(x_j)}{\partial a_m}$

$$\frac{\partial p(x_j)}{\partial a_m} = \phi' \left(\sum_{k=0}^{M-1} a_k z_k + d \right) z_m$$

$$\frac{\partial s_j}{\partial a_m} = 2[p(x_j) - y_j] \phi' \left(\sum_{k=0}^{M-1} a_k z_k + d \right) z_m$$

Similarly: $\frac{\partial s_j}{\partial b_m} = 2[p(x_j) - y_j] \phi' \left(\sum_{k=0}^{M-1} a_k z_k + d \right) a_m x_j \phi'(b_m x_j + c_m)$

$$\frac{\partial s_j}{\partial c_m} = 2[p(x_j) - y_j] \phi' \left(\sum_{k=0}^{M-1} a_k z_k + d \right) a_m \phi'(b_m x_j + c_m),$$

$$\frac{\partial s_j}{\partial d} = 2[p(x_j) - y_j] \phi' \left(\sum_{k=0}^{M-1} a_k z_k + d \right)$$

Neural network

Your task:

*Code this up, following the Unix philosophy
(write functions that do one thing and do it well)*

Needed components:

- Generate pseudodata
- Activation function
- Compute z 's
- Backpropagation
- Implement stochastic gradient descent

Acknowledgments

Funding

