

# Basic MLP with manually-derived Backprop

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## 1 Introduction

**Goal:** To design, train and use a simple 3-layer MLP for binary classification of size-2 vectors.

**Design:** of the form

$$[(layer\_size, Activation) \dots]$$

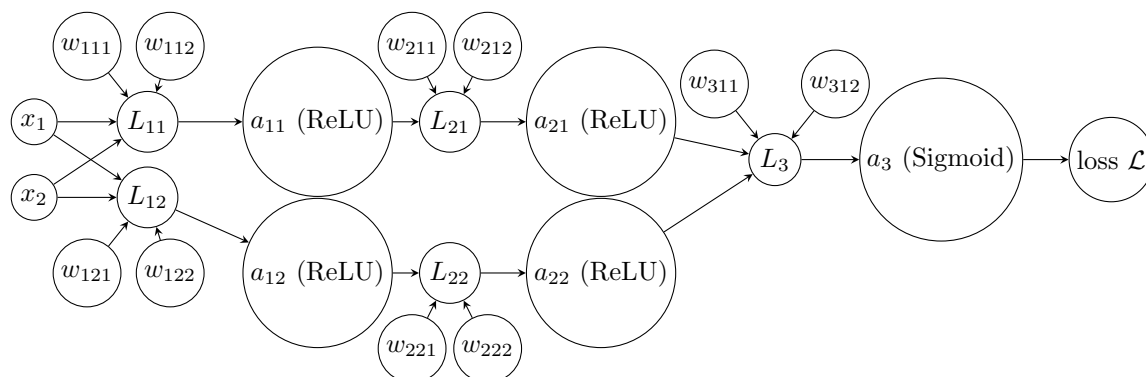
: [(2, ReLU), (2, ReLU), (1, Sigmoid)]

### 1.1 Diagrams

#### 1.1.1 Vectorized Diagram

$$\begin{array}{ccccccc} & \mathbf{w_1} & & \mathbf{w_2} & & \mathbf{w_3} & \\ & \searrow & & \searrow & & \searrow & \\ \mathbf{x} & \rightarrow \mathbf{L_1} & \rightarrow \mathbf{a_1} & \rightarrow \mathbf{L_2} & \rightarrow \mathbf{a_2} & \rightarrow \mathbf{L_3} & \rightarrow \mathbf{a_3} \rightarrow \mathcal{L} \end{array}$$

#### 1.1.2 Expanded Diagram



## 1.2 Definitions

### 1.2.1 Remark on weight notation

$w_{i,j,k}$  is to say the weight at the  $i$ -th layer,  $j$ -th neuron,  $k$ -th weight. Hence  $w_{111}$  is the first weight of the first neuron in the first layer, etc.

### 1.2.2 Remark on layer notation

This is a sub-case of the weight notation. I.e.,  $L_{ij}$  is the  $j$ -th neuron at the  $i$ -th layer, etc.

### 1.2.3 Neuron firing calculation

This is just a straightforward dot-product. We have:

$$L_i = \mathbf{w}_i \mathbf{x}_i$$

### 1.2.4

## 1.3 BackPropagation Derivation