Explanatory Notes for 6.390

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Fall 2022

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Many Layers

We are finally ready to build our **complete** neural network. We'll just retrace the steps of the 2-layer case.

Notation 1

The total number of layers in our neural network is notated as L.

Typically we notate an **arbitrary** layer as ℓ (or ℓ).

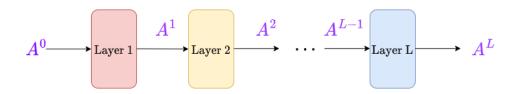
Since x is, for all purposes, **equivalent** to a vector A, we will call it A^0 .

Notation 2

Our **neural network**'s input x is used in the **same** way as every term A^{ℓ} .

So, we will **represent** it as

$$\boldsymbol{x} = \boldsymbol{A}^0$$



Again, we see that the **output** of layer ℓ is the **input** of layer $\ell + 1$.

Concept 3

Each layer **feeds** into the next layer.

 A^{ℓ} is the **output** of layer ℓ , and the **input** of layer $\ell + 1$.

This means that the output dimension must match the next input dimension.

$$\overbrace{n^\ell}^{Output} = \overbrace{m^{\ell+1}}^{Output}$$

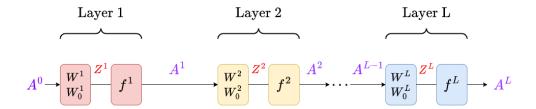
And the **dimension** of A^{ℓ} is $(n^{\ell} \times 1) = (m^{\ell+1} \times 1)$.

Our Complete Neural Network

We can break our layers into components, so we can see the functions involved.

With this, we build our final neural network:

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With this, we can see how each layer is **related** to each other: as we **mentioned**, the **output** of one layer is the **input** of the next layer.

Here is the computation we do for layer ℓ :

Key Equation 4

The calculations done by layer ℓ are given by

$$\mathsf{Z}^{\ell} = (\mathsf{W}^{\ell})^{\mathsf{T}} \mathsf{A}^{\ell-1} + \mathsf{W}_{0}^{\ell}$$

and

$$A^{\ell} = f(Z^{\ell})$$

Which combine into:

$$A^{\ell} = f(Z^{\ell}) = f\bigg((W^{\ell})^{\mathsf{T}} A^{\ell-1} + W_0^{\ell}\bigg)$$

One more comment: a useful definition.

Definition 5

A hidden layer is any layer except for the last one.

It is called a "hidden" layer because, if you're viewing the whole neural network based on

- Input x (first input)
- Output A^L (final output)

Then you can't see the **output** of any of the layers except for the **last** one.

Sometimes you'll hear someone say that a hidden layer is any except the "first or last": by that, they mean you can view the input for the first layer, as well as the output for the last layer.

But, when we're talking about **activation** functions (which we often are when we mention hidden layers, see below), we only care about whether the **output** is hidden!