

# Chain Rule in Back Propagation

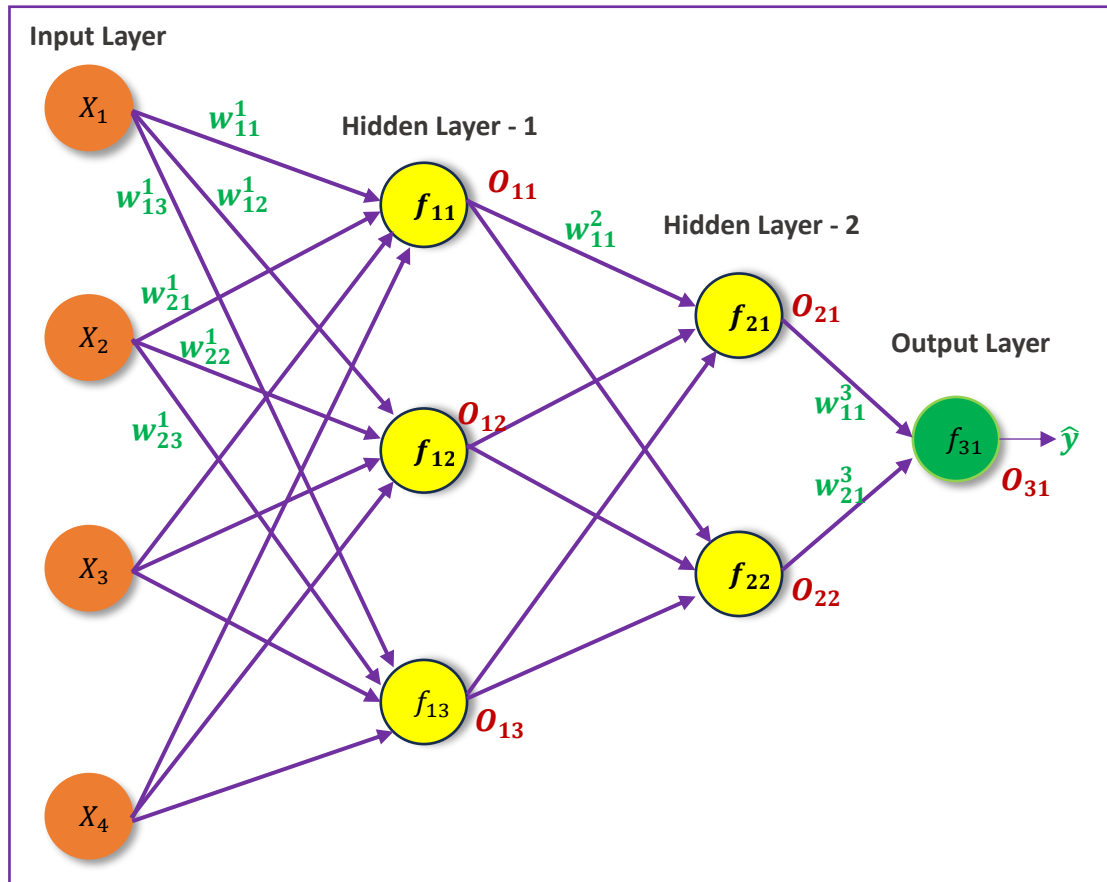


Figure 1: Multilayer neural network

Let's take gradient descent wide updating formula for reference:

$$\omega_{new} = \omega_{old} - \alpha \frac{\partial L}{\partial \omega_{old}} \dots \dots \dots (1)$$

The chain rule formula for update  $w^3_{11}$  at first step from output side is:

$$w^3_{11new} = w^3_{11} - \alpha \frac{\partial L}{\partial w^3_{11}} \dots \dots \dots (2)$$

$$\frac{\partial L}{\partial w_{11}^3} = \frac{\partial L}{\partial O_{31}} \cdot \frac{\partial O_{31}}{\partial w_{11}^3}$$

We should calculate  $\frac{\partial L}{\partial w_{11}^3}$  by using above formula and substitute in  $w_{11new}^3$  (equation- 2)

Similarly, the formula for update the  $w_{21}^3$  is:

$$w_{21new}^3 = w_{21}^3 - \alpha \frac{\partial L}{\partial w_{21}^3} \dots \dots \dots (3)$$

$$\frac{\partial L}{\partial w_{21}^3} = \frac{\partial L}{\partial O_{31}} \cdot \frac{\partial O_{31}}{\partial w_{21}^3}$$

The formula for update the  $w_{12}^2$  is:

$$w_{12new}^2 = w_{12}^2 - \alpha \frac{\partial L}{\partial w_{12}^2} \dots \dots \dots (2)$$

$$\frac{\partial L}{\partial w_{12}^2} = \left[ \frac{\partial L}{\partial O_{31}} \cdot \frac{\partial O_{31}}{\partial O_{21}} \cdot \frac{\partial O_{21}}{\partial w_{12}^2} \right] + \left[ \frac{\partial L}{\partial O_{31}} \cdot \frac{\partial O_{31}}{\partial O_{22}} \cdot \frac{\partial O_{22}}{\partial w_{12}^2} \right]$$

It will absorb the above equation The second term is extra added after + Because when we want to change the weight  $w_{12}^2$  the output of  $O_{31}$  affected in two paths are  $w_{11}^2 \rightarrow O_{21} \rightarrow O_{31}$  and  $w_{12}^2 \rightarrow O_{22} \rightarrow O_{31}$ . That's why second chain added in above equation.

Note: This is just a basic chain calculation, if you assume 100 connections of neurons are existed, then the formula will be bigger and many differentiations will be performed in that formula.