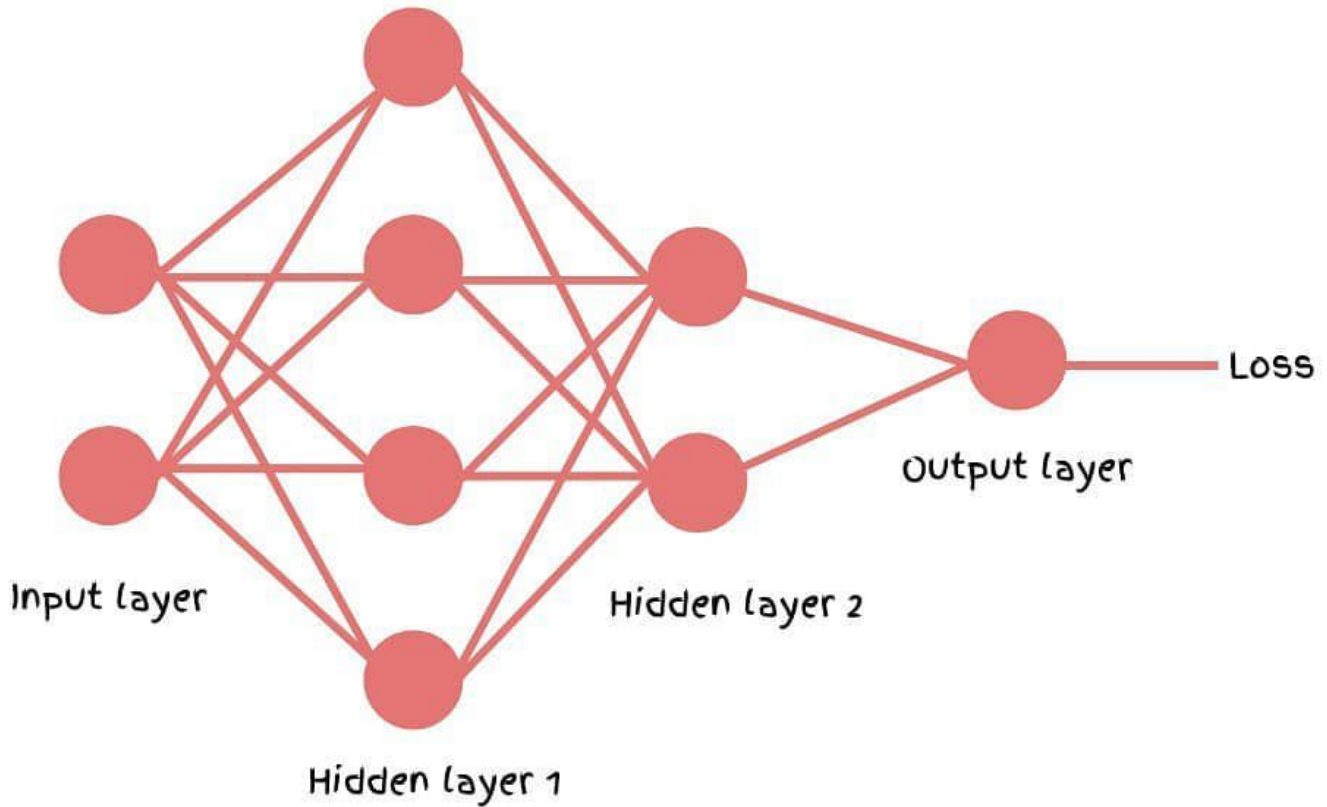
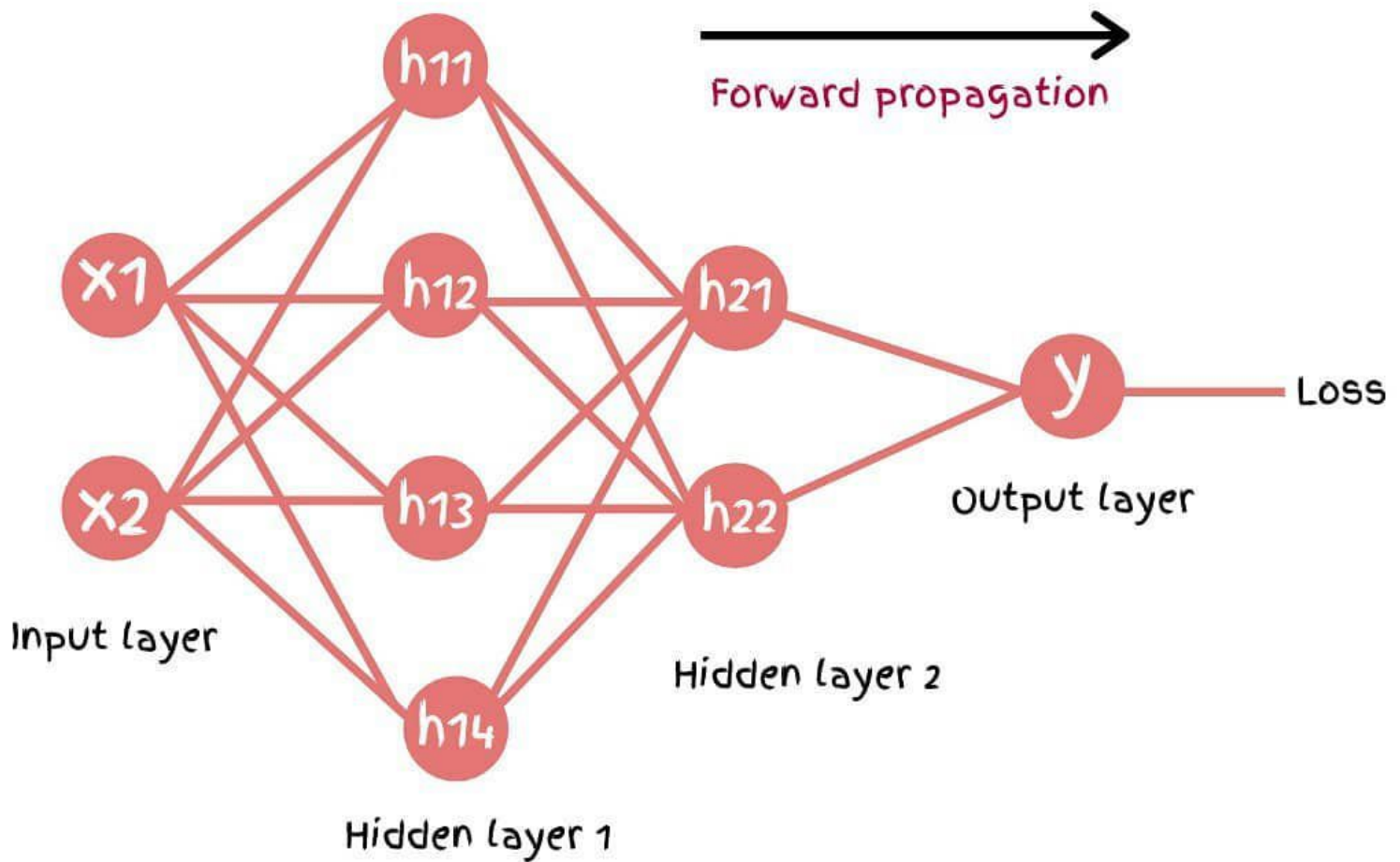


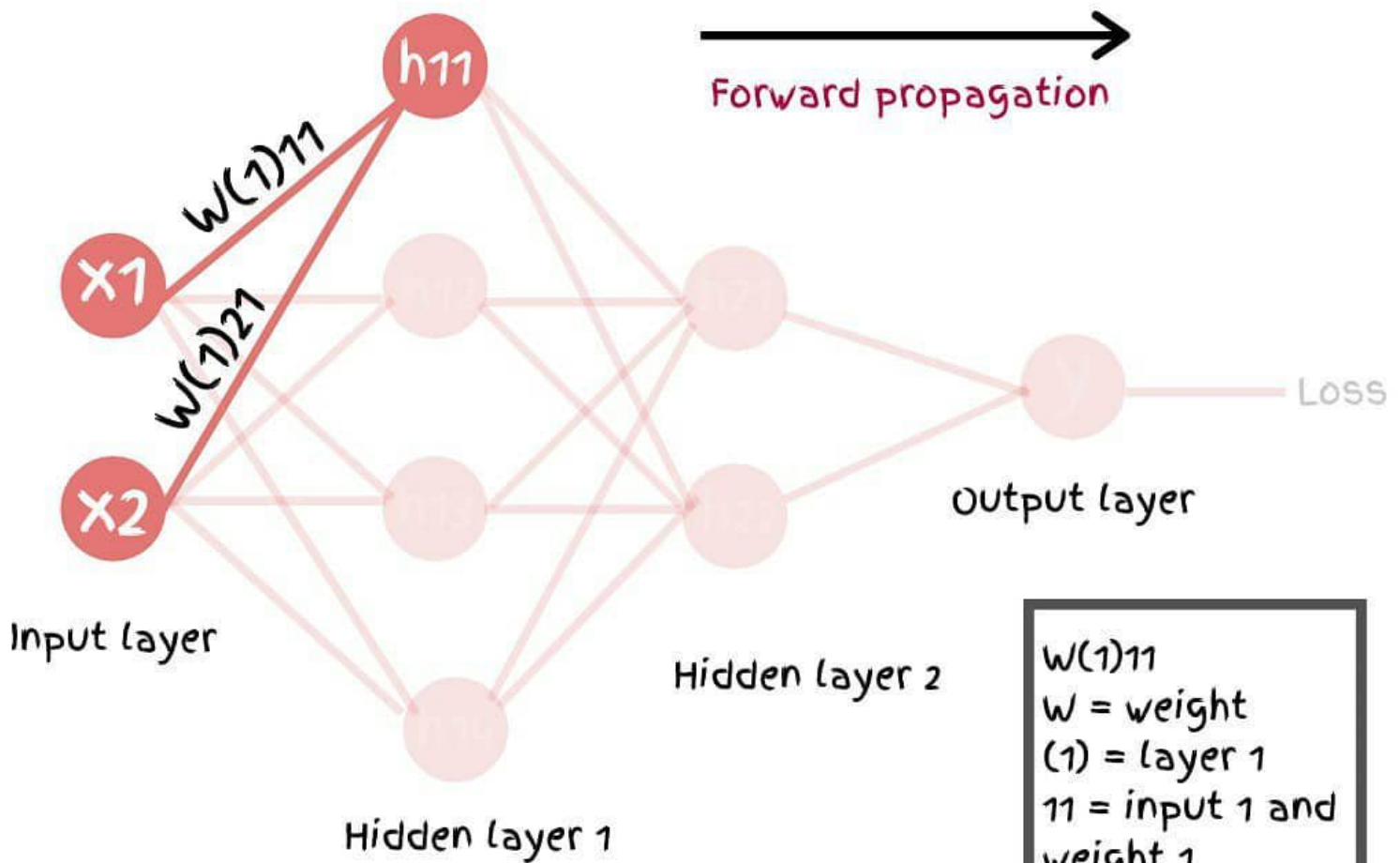
→ Forward propagation



How feed-forward
propagation works??



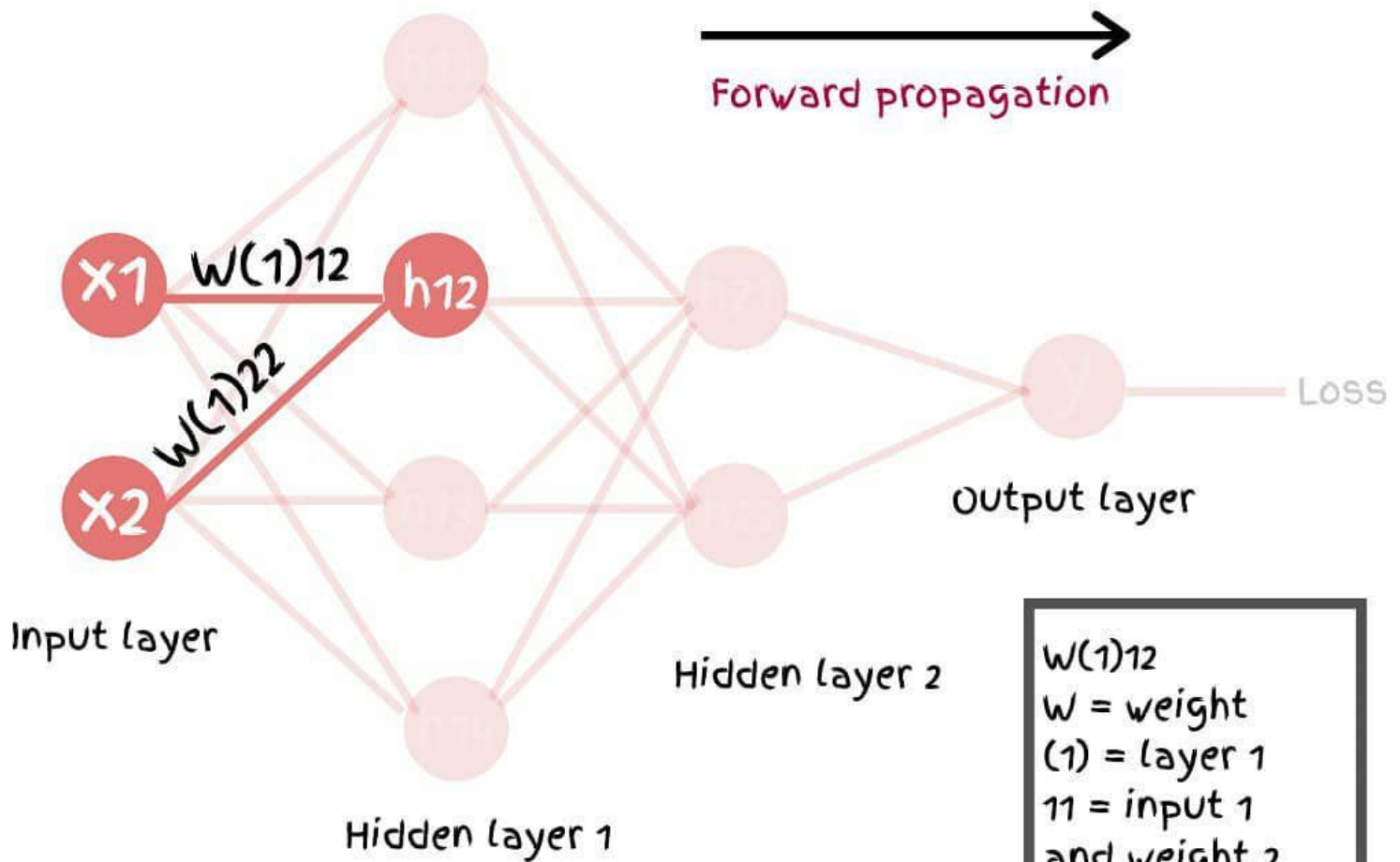
- So here we have 2 features x_1 and x_2 .
- And we have 2 hidden layers h_1 and h_2 .
- Each hidden layer has few neurons like we have 4 neurons in hidden layer h_1 and named as $\{h_{11}, h_{12}, h_{13}, h_{14}\}$ and for second hidden layer h_2 $\{h_{21}, h_{22}\}$



$w^{(1)11}$
 w = weight
 (1) = layer 1
 11 = input 1 and weight 1

$$h_{11} = f(x_1 * w^{(1)11} + x_2 * w^{(1)21})$$

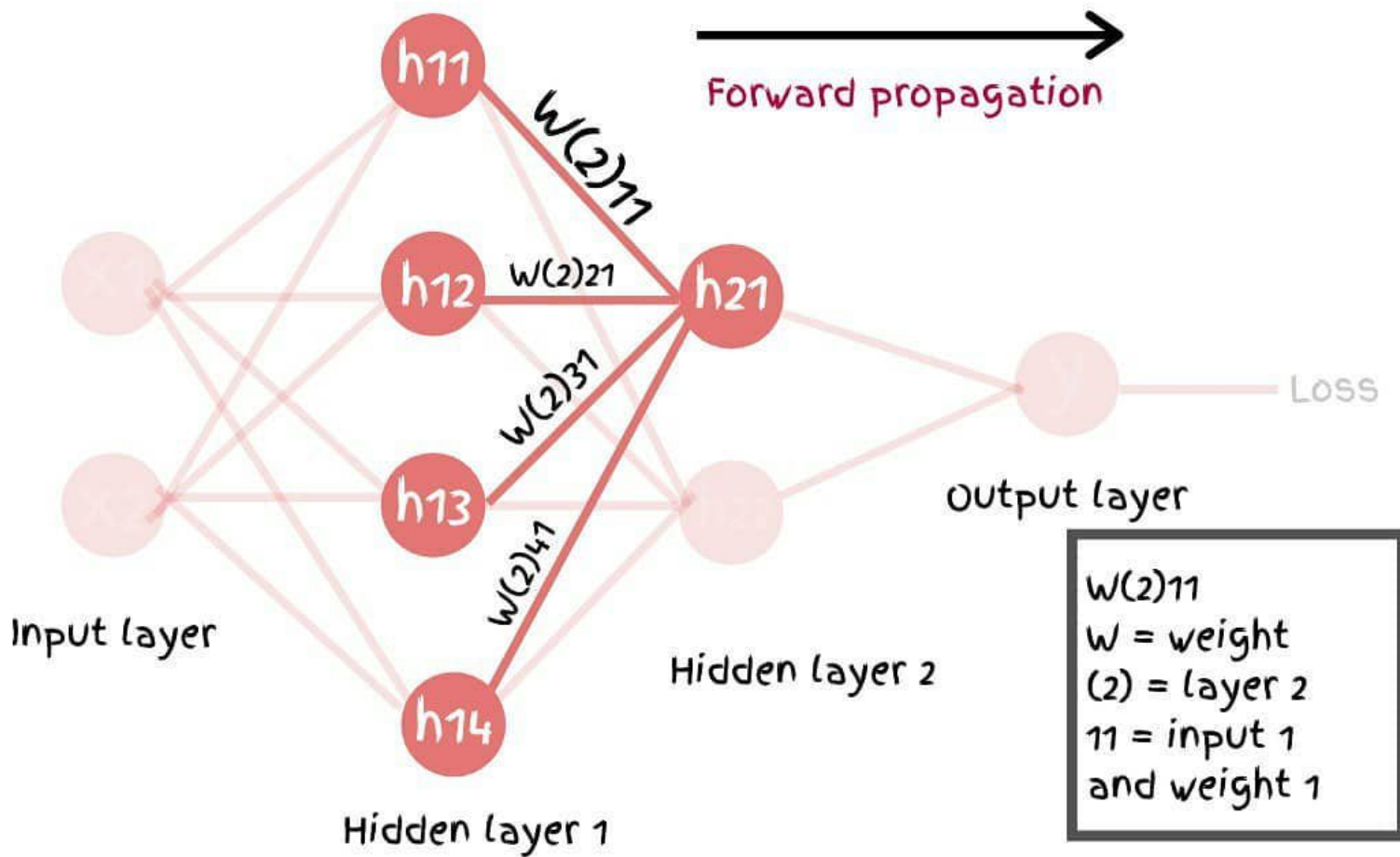
- The first operation we will be doing is to calculate h_{11} .
- We will be doing the dot product of input with the respective weights associated with that neuron.
- Ans we will apply activation function on the dot product.



$w^{(1)}_{12}$
 w = weight
 (1) = layer 1
 11 = input 1
 and weight 2

$$h_{12} = f(x_1 * w^{(1)}_{12} + x_2 * w^{(1)}_{22})$$

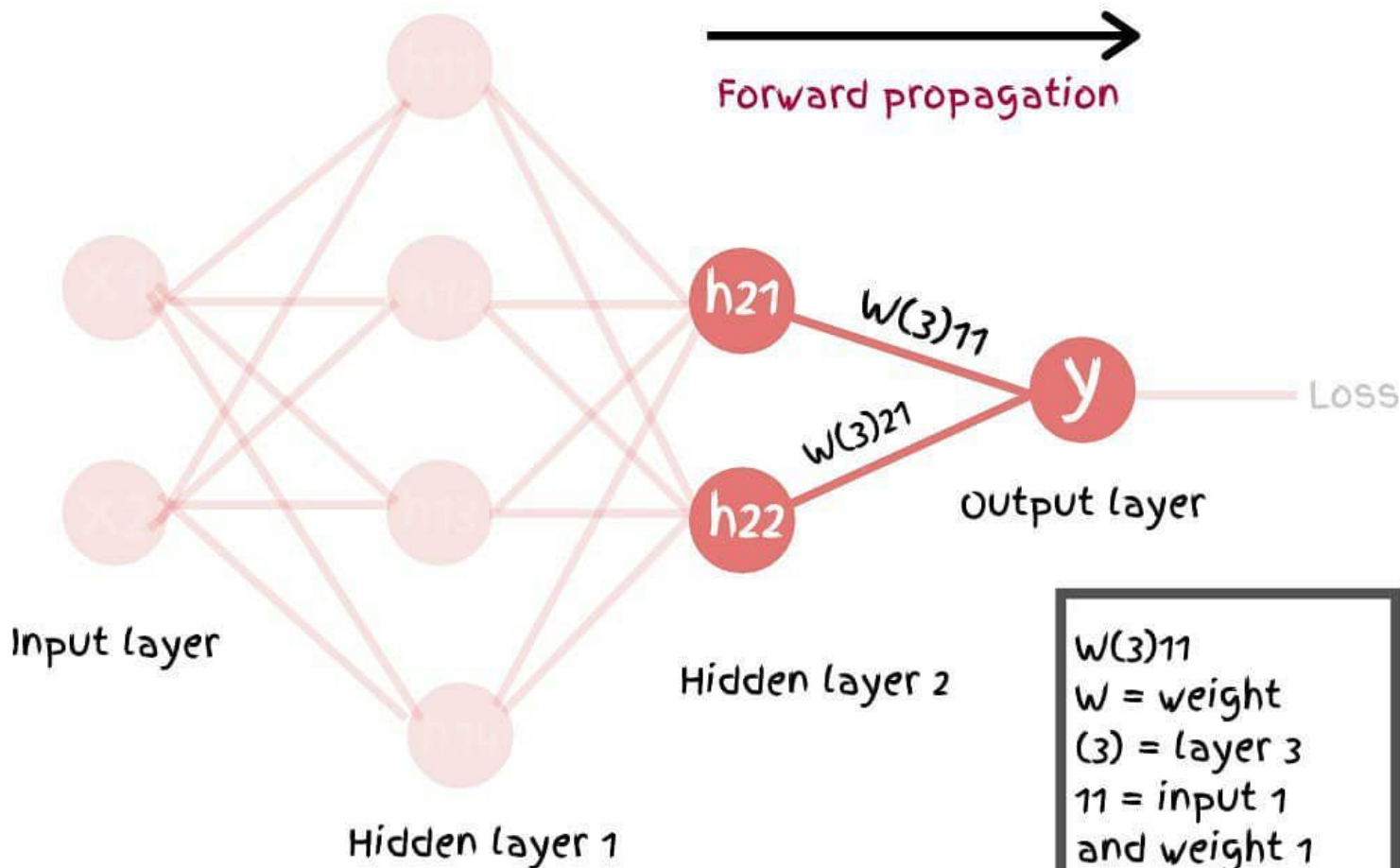
- The first operation we will be doing is to calculate h_{12} .
- We will be doing the dot product of input with the respective weights associated with that neuron.
- Ans we will apply activation function on the dot product.



$W(2)_{11}$
 W = weight
 (2) = layer 2
 11 = input 1
 and weight 1

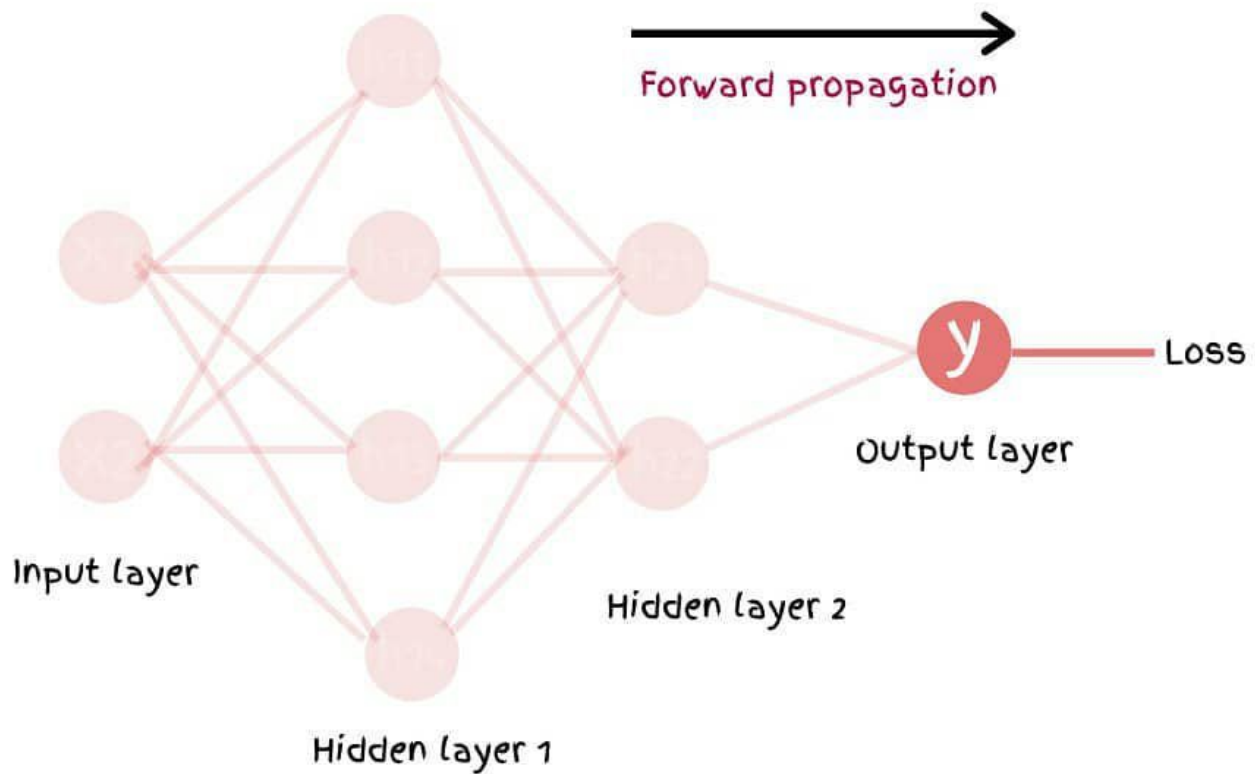
$$h_{21} = f(h_{11} * W(2)_{11} + h_{12} * W(2)_{21} + h_{13} * W(2)_{31} + h_{14} * W(2)_{41})$$

- The first operation we will be doing is to calculate h_{21} .
- We will be doing the dot product of input with the respective weights associated with that neuron.
- Ans we will apply activation function on the dot product.



$$y = f(h_{21} * W(3)_{11} + h_{22} * W(3)_{21})$$

- The first operation we will be doing is to calculate y .
- We will be doing the dot product of input with the respective weights associated with that neuron.
- Ans we will apply activation function on the dot product.



- After Y is calculated then it is compared with ground truth label and the loss is calculated
- Let it be regression problem and loss function is MSE

$$\text{Loss} = (1/n) \sum (y - y_{\text{orig}})^2$$

- Let it be binary classification problem. so the output of that y will be a probability and loss function is log loss

$$\text{Loss} = - (1/n) \sum (y_{\text{orig}} * \log(y) + (1 - y_{\text{orig}}) * \log(1 - y))$$