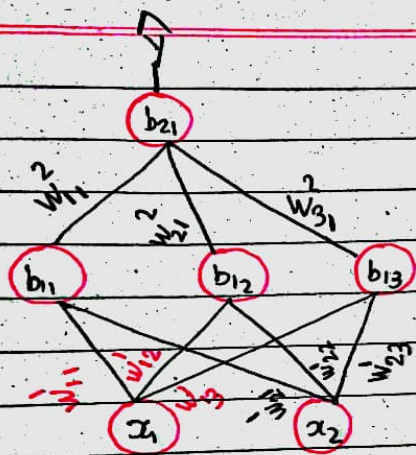


Dropout layers

Suppose hami sanga ya neural Network ka Tasle
euta classification problem solve garxa.



output layer = 1

1st hidden layer = 3

Input layer = 2

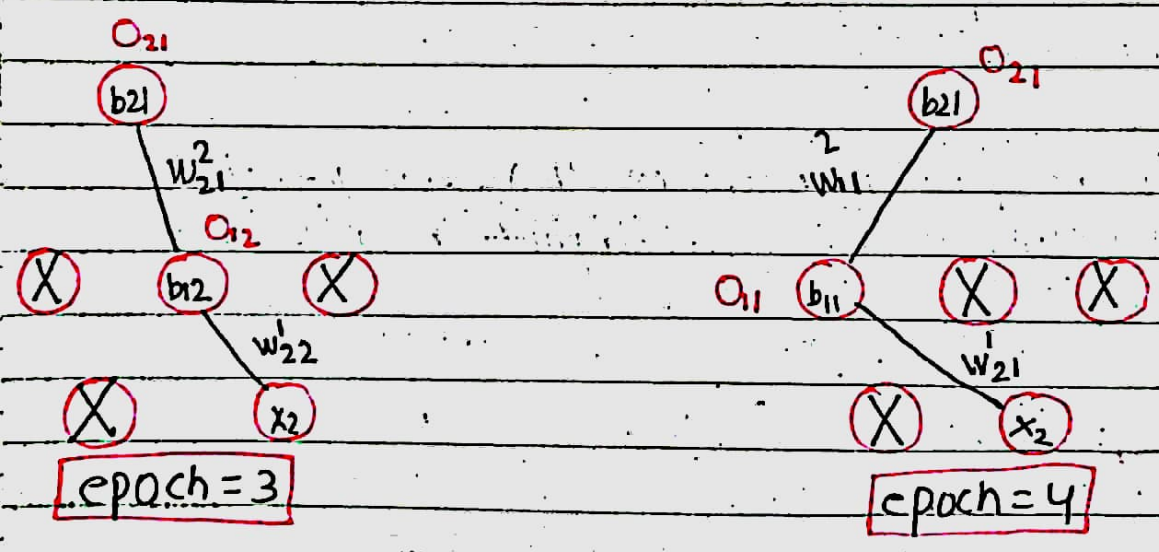
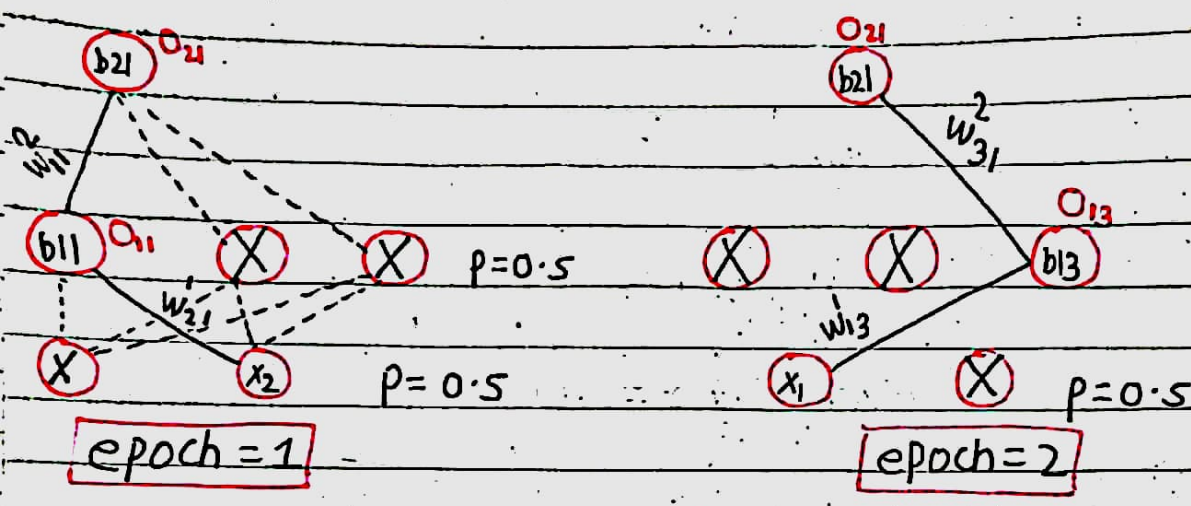
fully connected neural
network

When we use this complex Architecture there will be the
problem of the overfitting so, we use dropout in order to
solve this problem

The concept of dropout:

Suppose hami afno neural network **epoch = 4** ko lagi
train gardai xam. Jaha hamiley dropout use garxam
hamro neural network ma It will **randomly switch off**
Some of the nodes of **input and hidden layer** in each
epochs.

aba hamo neural network ko architecture herek epoch ko lag? different hunxa like down:-



Now, we have dataset as:

Age	purchase	Salary
5	0	0
20	1	200
10	0	50
30	1	70

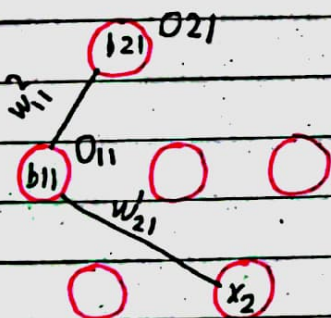
Suppose we took batch = 2

$$\text{Batch size} = \frac{\text{Total row}}{\text{batch}} = \frac{4}{2} = 2$$

x_1	x_2	y	x_1	x_2	y
Age	Purchase	Sal	Age	Purchase	Sal
20	1	200	5	0	0
30	1	70	10	0	50

At first, we initialize the parameters (All) of our neural network.

At epoch 1



During Forward propagation, we calculate

$$O_{11} = x_2 \cdot w_{21} + b_{11}$$

$$O_{21} = O_{11} \cdot w_{11} + b_{21}$$

(Batch 1 ko first row ko x_2 feature Tanxa calculate hunxa, 2nd row ko x_2 feature Tanxa feri calculate hunxa, Sabai row sake paxi batch ko loss calculate hunxa ani back propagation vai (w_{21} , w_{11} , b_{11} , b_{21}) update hunxa.)

(Abu bhakar updated weights ma arko batch 2 ko dubai row ko x_2 feature NN ma Tanxa, loss calculate hunxa feri weights update hunxa)

Let's see how weight update takes place:
 (Suppose we have linear activations to all nodes)

$$w_{11}^2_{\text{new}} = w_{11}^2_{\text{old}} - \alpha \frac{d\text{loss}}{dw_{11}^2_{\text{old}}}$$

A

$$[w_{11}^2 \rightarrow O_{21} \rightarrow \text{loss}]$$

$$\frac{d\text{loss}}{dw_{11}^2} = \frac{d\text{loss}}{dO_{21}} \times \frac{dO_{21}}{dw_{11}^2}$$

$$= \left\{ \sum_{i=1}^2 \frac{d(y_i - \hat{y}_i)^2}{d(y - \hat{y})} \times \frac{d(y - \hat{y})}{dO_{21}} \right\} \times \frac{dO_{21}}{dw_{11}^2}$$

$$= \left\{ \sum_{i=1}^2 2(y_i - \hat{y}_i) \right\} \times (0 - 1) \times \frac{d[O_{21} \cdot w_{11}^2 + b_{21}]}{dw_{11}^2}$$

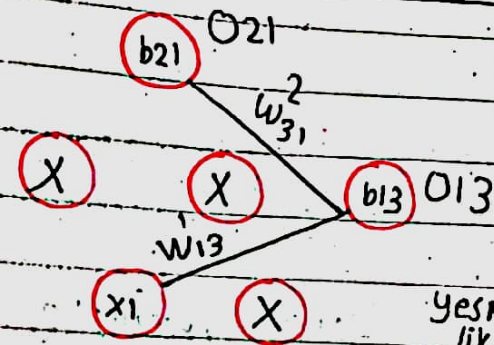
$$= \left\{ \sum_{i=1}^2 2(\hat{y}_i - y_i) \right\} \times O_{11}$$

$$= 2[(\hat{y}_1 - y_1) + (\hat{y}_2 - y_2)] \times O_{11}$$

$$= 2[(\hat{y}_1 - y_1) + (\hat{y}_2 - y_2)] \times [x_2 \cdot w_{21}^1 + b_{11}]$$

$$\therefore w_{11}^2_{\text{new}} = w_{11}^2_{\text{old}} - \alpha \times [2[(\hat{y}_1 - y_1) + (\hat{y}_2 - y_2)] \times [x_2 \cdot w_{21}^1 + b_{11}]]$$

Similarity for epoch 2



During forward propagation, we calculate

$$O_{13} = x_1 \cdot w_{13} + b_{13}$$

$$O_{21} = O_{13} \cdot w_{31} + b_{21}$$

Yesma last updated like b_{21}

weight xa vane tesmai train huni vayo

Similarly yesma pani agi jastai hunxa.
Batch 1 ko row 1 bata x_1 feature
Janxa, tespari row 2 bata x_1 feature
Janxa. dubai ko \hat{y} palai pila calculate
hunxa. loss calculate hunxa, backpropagation
ma parameters (b_{21} , b_{13} , w_{31} , w_{13})
update hunxa)

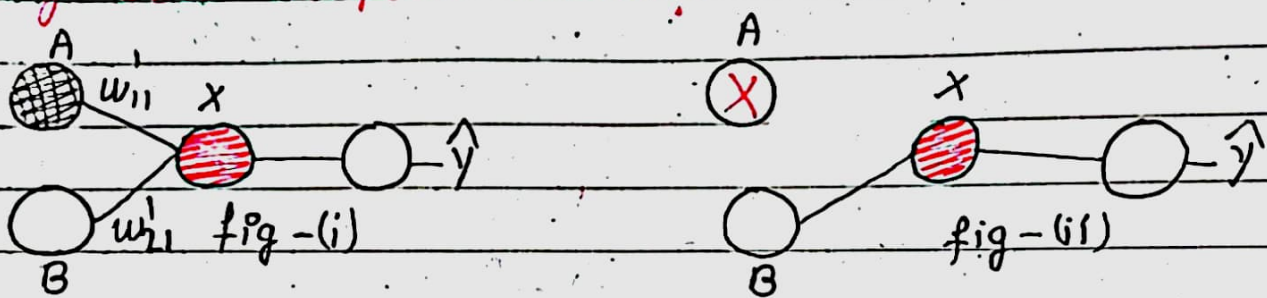
(feri bhakar update vaye ko parameters
ma batch 2 ko data train garinxa,
 \hat{y} garinxa, loss calculate garinxa and
feri back propagation use gari parameter)
update garinxa)

Yesari harek epoch ma naya naya architecture form hunxa
kunai node off kunai on vayera. ani weights pani
Sabai update vai rako hunxa

Dropout ratio (p):

Suppose I choose dropout ratio as 0.5 i.e. $p=0.5$. yesko matlab yesle randomly harek epoch ma 50% neuron lai off gardinxo.

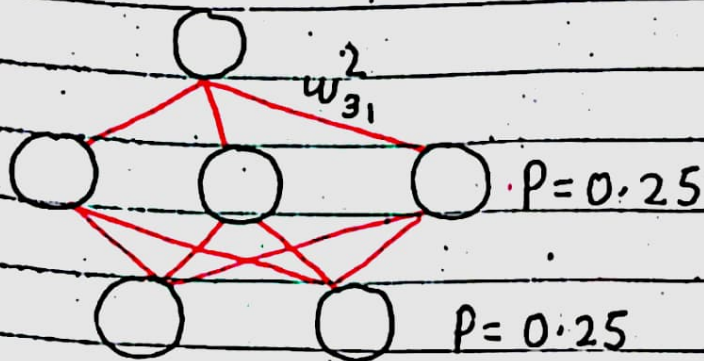
Why does dropout works?



A = no of rooms, B = no of gardens

When we feed our data to this NN. Node X more focuses on feature A . Since, no of rooms have strong correlation **to the** price of house than no of gardens. due to this reason Node X assign higher weights value for w_{11} and due to this reason there may be overfitting problem because it won't give priority to feature B and can't learn proper pattern. but when we we dropout as in figure (ii) the node X also look for feature B and also give time to learn patterns from feature B which prevents w from overfitting.

How prediction happens?



This dropout only happens during training. Jaba Sabai data train vayera hami ley aho best weights pavxam. Tyo weight lai nai we gardenam hamle. testing ko lagi

Testing ko time ma Sabai Node available hunxy. So, what will be the weights value? It was not guaranteed the w_{31}^2 was available or not ^{during training} (it was taken as an example it also worked for others). Suppose the value we got w_{31}^2 is w . Now, for testing $w_{31}^2 = w(1-p)$
 $= w \times 0.75$

Why this $w \times 0.75$?

⇒ yoo w_{31}^2 training ko time ma active hune probability 0.75 matra xa, 0.25 probability ma ^{1 layer node 3 bata kunai data na gako huna skxa} ^{tei} vara hami training time ma active hune probability sanga multiply garera testing garxam

(This all is handled by keras)