

$$(2) \text{ Expected utility of playing} = 0.1(100 \times 0.8 + 50 \times 0.2) + 0.9(-50 \times 0.2 + 0 \times 0.8)$$

$$= 0.1(90) + 0.9(-10)$$

$$= 9 - 9 = 0$$

$$\therefore \text{ Expected utility of playing} = 0$$

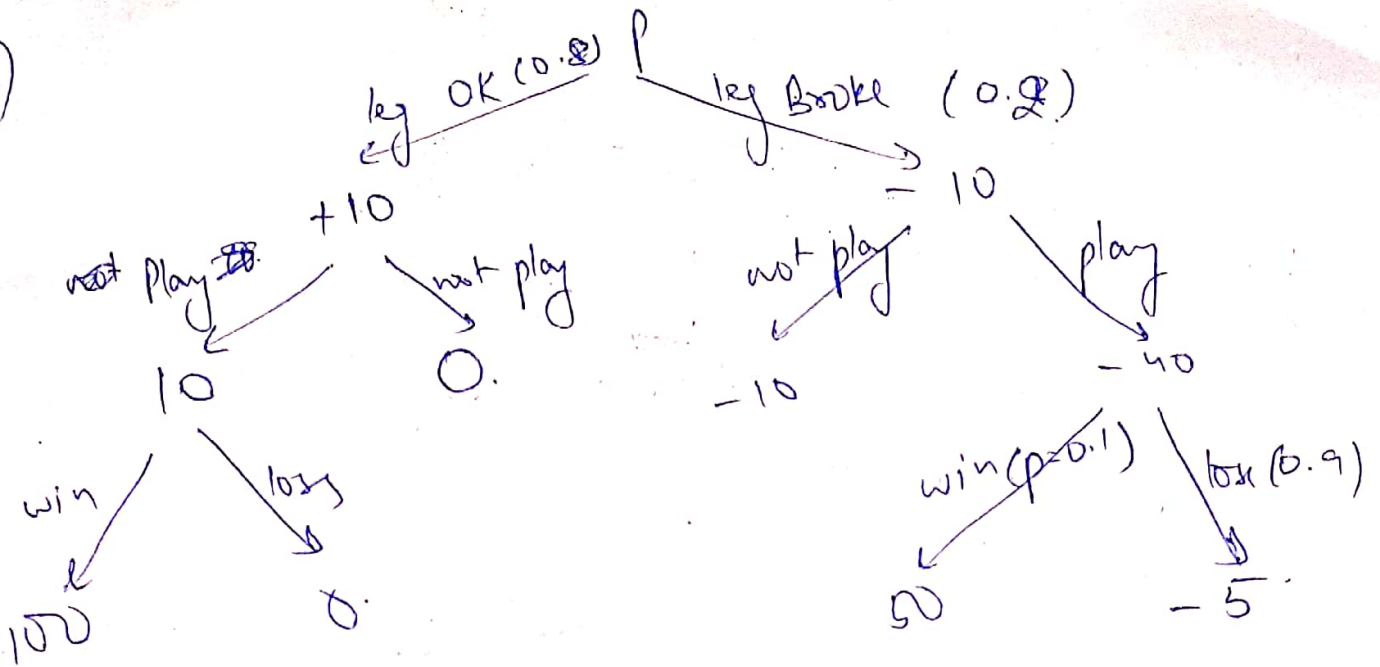
$$\text{ Expected utility of not playing} = 0.2(-10) + 0.8(0)$$

$$= -2$$

$$\therefore \text{ Expected utility of not playing} = -2$$

Since $0 > -2$, \therefore I should play.

A3)

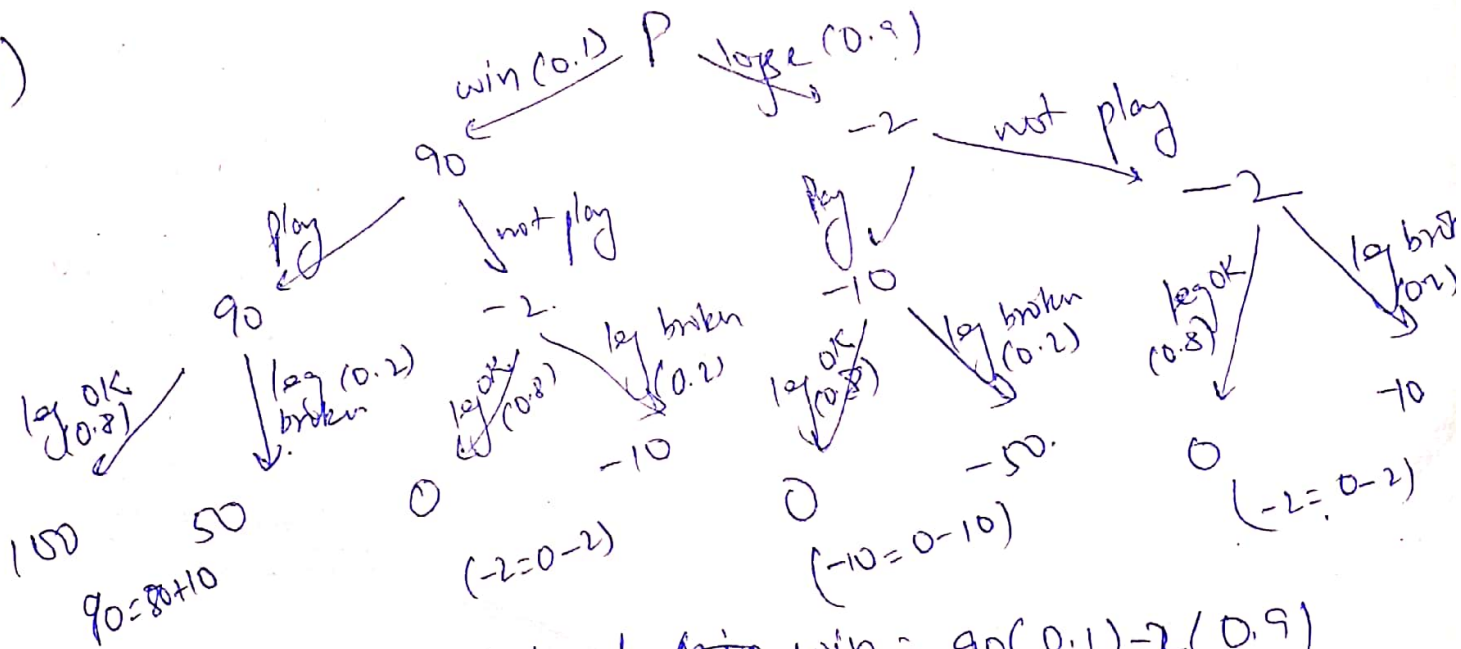


~~Expected info~~

Expected value perfect info of the leg =

$$0.2 \times 10 + 0.8 \times 10 = -2 + 8 = 6$$

4)



∴ Expected val of info of ~~fair~~ win = $90(0.1) - 2(0.9)$
 $= 9 - 1.8$
 $= 7.2$

(5) To find $P[\text{win} | \text{broken}]$

If winning decision branch is marked after LEG OK & LEG BROKEN, the reqd prob can be calculated & that can help to see whether win is more likely or loss.

Question 2.)

T.P. $f: \{0,1\}^d \rightarrow \{0,1\}$

can be represented as a neural network with just 1 hidden layer.

Now this question can be seen like having a black box which takes any number of inputs & outputs either 0 or a 1.

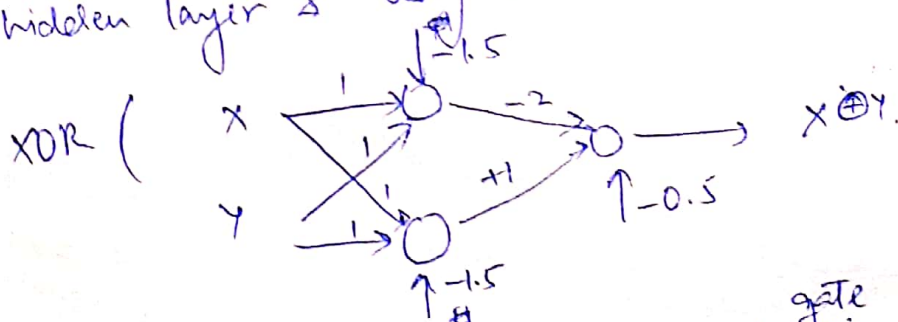
So this question can be modelled as proving that we can present a ~~single hidden layer~~ ~~perceptron~~ boolean function as a single layer with the help of a single ~~hidden~~ hidden layer.

Now we know that simple gates like

OR $\left(\begin{matrix} x \\ y \end{matrix} \right) \xrightarrow{2} \bigcirc \xrightarrow{-1} x \vee y$ AND $\left(\begin{matrix} x \\ y \end{matrix} \right) \xrightarrow{1} \bigcirc \xrightarrow{-1} x \wedge y$

NOT $\left(x \xrightarrow{-1} \bigcirc \xrightarrow{-1} \bar{x} \right)$ can be represented directly as single perceptrons.

When we need functions like XOR, we introduce a hidden layer & use that XOR can be easily implemented.



Now we know that any ~~function~~ ^{gate} function can be created using AND and NOT gates (from ECE 101 (Digital ckt) (omitting the proof work))

∴ we can reduce any expression to a series of and and or gates. Thus we can use these gates to model any boolean logic (part of hidden layer) & their output would be a boolean value either 0 or 1.

∴ This can be illustrated using XOR gate itself on the previous page.

Thus since any boolean function can be represented using AND and OR gates & these can be implemented using just a single perceptron,

∴ Any boolean function can be implemented using just ~~the~~ a single hidden layer.

Ans 3)

2 input nodes, 1 hidden layer

$$Y = \arg\max_y P(Y=y | X)$$

$$Y = P(Y=1 | X) = \frac{\exp(\beta_1 X_1 + \beta_2 X_2)}{1 + \exp(\beta_1 X_1 + \beta_2 X_2)}$$

$$Y = P(Y=-1 | X) =$$

