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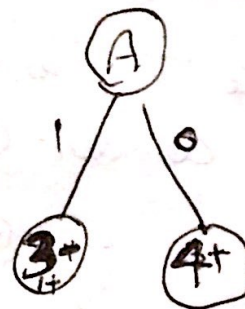
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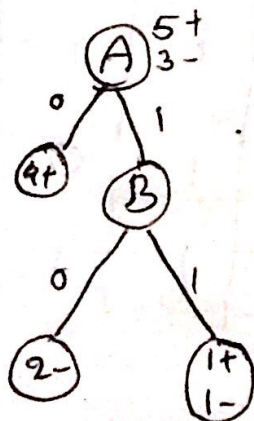
① (a)  $Y = (\neg A \vee B) \wedge \neg(C \wedge A)$  | A, B, C are boolean  
 Truth table:

A	B	C	Y
1	1	1	0
1	1	0	1
1	0	1	0
0	1	1	1
1	0	0	0
0	0	1	1
0	1	0	1
0	0	0	1

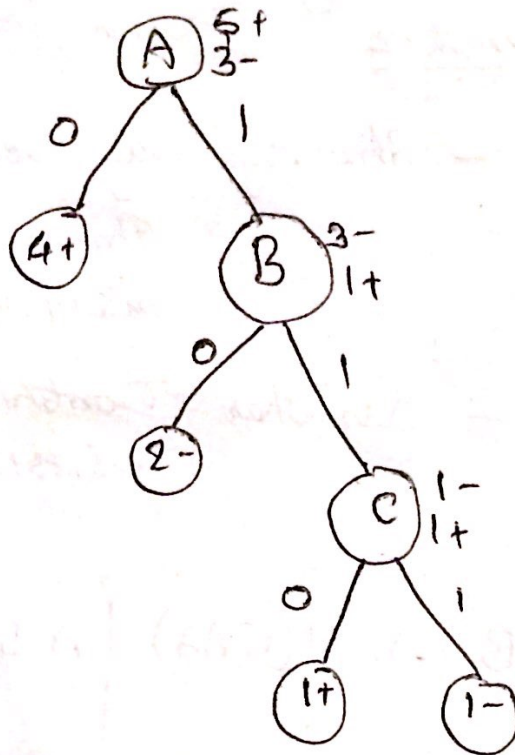
Step ①  
 since  $\forall$  values of  
 (0) A, there is value  
 1 in Y, consider A  
 as root node,



Step ② split wrt B



step 2 split w.r.t. C :-



This is final Decision Tree.

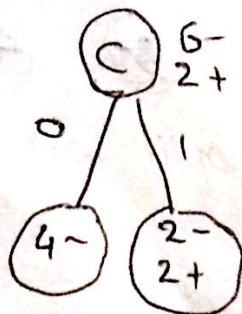
(b)  $(A \oplus B) \wedge C$

$\Downarrow$

$Y = (A'B \vee B'A) \wedge C$

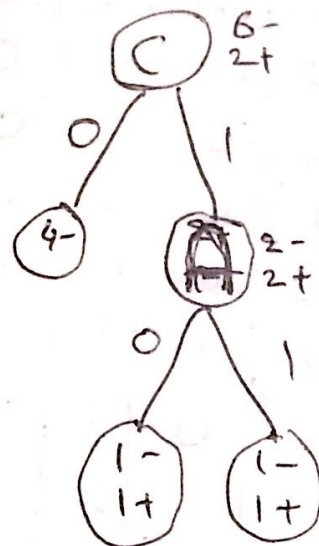
step 1

Since  $\forall C=0 \Rightarrow Y=0$ ,  
so consider C as root,

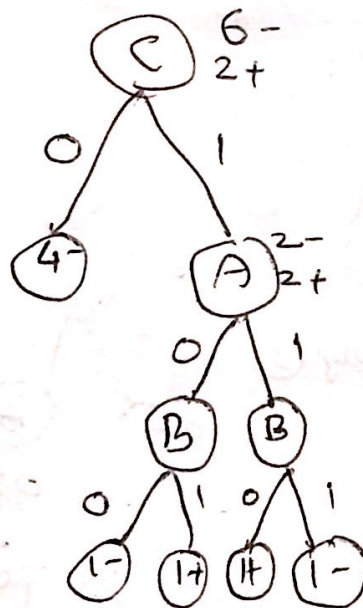


A	B	C	Y
0	0	0	0
0	0	1	0
0	1	0	0
1	0	0	0
0	1	1	1
1	1	0	0
1	0	1	1
1	1	1	0

step ② Consider A as next node, when  $C=1$  because  $C=0$  is pure.



step ③, since A has equal node values, consider B for each of children nodes of A.



This is final decision tree

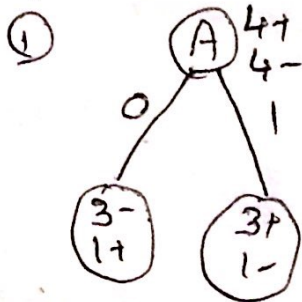


$$(c) Y = (A \vee B) \wedge (B \vee C) \wedge (A \vee C)$$

Truth Table:

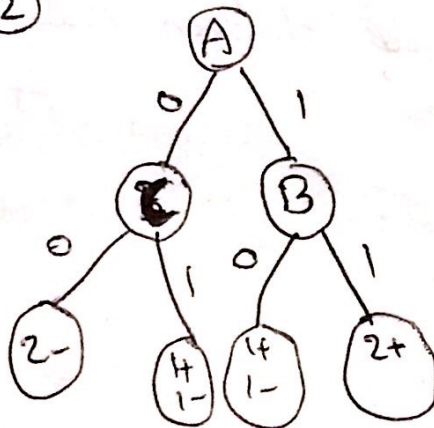
Since here, the Y is symmetric, consider any feature as root nodes

A	B	C	Y
0	0	0	0
0	0	1	0
1	0	0	0
0	1	0	0
0	1	1	1
1	0	1	1
1	1	0	1
1	1	1	1

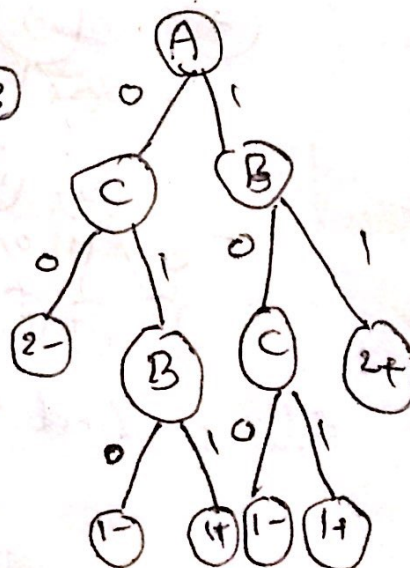


cost B

step ②



step ③



This is final decision Tree

$$(d) Y = (A \vee B) \wedge \neg A \wedge \neg B$$

Truth Table:



A	B	Y
0	0	0
0	0	0
0	1	0
1	0	0
1	0	0
1	1	0
0	1	0
1	1	0

$$(d) Y = A \vee B \wedge \neg A \wedge \neg B$$



A	B	Y
0	0	0
0	1	0
1	0	0
1	1	0

Since,  $Y = 0$   $\forall$  values of  $A \in B$   
 both  $A$  &  $B$  can be chosen as root node  
 with no further classification needed.

(A) or (B) is Final decision Tree.

(2)

Instance	$X_1$	$X_2$	$X_3$	Class
1	1	0	0	1
2	0	1	0	0
3	0	0	0	0
4	1	0	1	0
5	0	0	0	0
6	1	1	0	1
7	0	1	1	0
8	1	0	0	1
9	0	0	0	0
10	1	0	0	1

According to ID3, check Information gain for the attributes  $X_1, X_2, X_3$ .  
Feature with Highest I.G. is root Node.

Information Gain,  $I.G. = \text{Entropy (parent)} - \text{avg. entropy (children)}$

$$\text{Entropy}(E) = \sum_{i=1}^n -P_i \log P_i$$

Number of +ve(1) in Class = 5

-ve(0) in class = 5

$$E_{\text{parent}} = - \left[ \frac{5}{10} \log \frac{5}{10} + \frac{5}{10} \log \frac{5}{10} \right]$$

$$= +1$$

Entropy of  $X_1 =$

$$H(Y/X_1) = - \sum P(X_1) \sum P(Y/X_1) \log P(Y/X_1)$$

$$= - \sum_{X_1} P(X_1) \sum_Y H(Y/X=X_1)$$



$$H(\text{Class}/X_1=\infty) = - \left[ \frac{4}{5} \log_2 \frac{4}{5} + \frac{1}{5} \log_2 \frac{1}{5} \right]$$

$$= 0.721928$$

$$H(\text{Class}/X_1=1) = - \left[ \frac{4}{5} \log_2 \frac{4}{5} + \frac{1}{5} \log_2 \frac{1}{5} \right]$$

$$= 0.721928$$

$$IG = 1 - H(\text{Class}/X_1)$$

$$= 1 - \left[ \frac{5}{10} \times 0.7219 + \frac{5}{10} \times 0.7219 \right]$$

$$= 1 - 0.7219 = 0.278$$

$$\text{Entropy}(X_2) =$$

$$H(Y/X_2) = - \sum_{x_2} P(x_2) \sum_y H(Y/x=x_2)$$

$$H(\text{Class}/X_2=\infty) = - \frac{4}{7} \log_2 \frac{4}{7} - \frac{3}{7} \log_2 \frac{3}{7}$$

$$= 0.9849$$

$$H(\text{Class}/X_2=1) = - \frac{2}{3} \log_2 \frac{2}{3} - \frac{1}{3} \log_2 \frac{1}{3}$$

$$= 0.9182$$

$$H(\text{Class}/X_2) = \frac{7}{10} \times (0.985) + \frac{3}{10} \times (0.918)$$

$$= 0.965$$

$$IG = 1 - 0.965 = 0.035$$

Entropy of  $X_3$

$$H(Y|X_3=0) = -\frac{3}{8} \log \frac{3}{8} - \frac{5}{8} \log \frac{5}{8}$$

$$= 0.9543$$

$$H(Y|X_3=1) = -\left[\frac{0}{2} \log \frac{0}{2} + \frac{2}{2} \log \frac{2}{2}\right] = 0$$

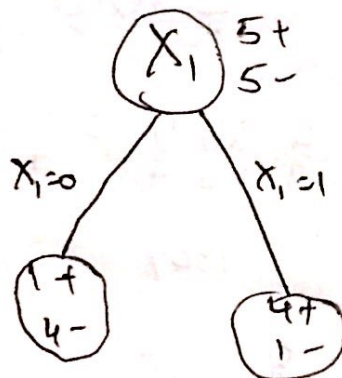
$$H(\text{Class}/X_3) = 0.9543 \times \frac{8}{16}$$

$$= 0.47635$$

$$IG = 1 - 0.47635$$

$$= 0.52365$$

$\Rightarrow IG(X_1)$  is highest, so  $X_1$  is root node.



Here, we have to decide  $X_2/X_3$  as our left or right nodes based on the information gain ( $X_1=0$  &  $X_1=1$  do not give pure nodes)

For Left nodes:-

Entropy<sub>when  $X_1=0$</sub> ( $X_1$ ) = 0

$$H(X_1=0/X_2=0) = -\left[\frac{3}{3} \log \frac{3}{3}\right] = 0$$

$$H(X_1=0/X_2=1) = -\left[\frac{1}{2} \log \frac{1}{2} + \frac{1}{2} \log \frac{1}{2}\right]$$

$$= 1$$



$$\text{Entropy } H(X_1=0/X_2) = \frac{3}{5} \times 0 + \frac{2}{5} \times 1$$

$$= 0.4$$

$$\therefore IG = 0.3219$$

$$\left( X_2 \text{ w.r.t } X_1=0 \right)$$

(left side)

Entropy of  $X_3$  when  $X_1=0$

$$\Rightarrow H(X_1=0/X_3=0) = - \left[ \frac{1}{4} \log \frac{1}{4} + \frac{3}{4} \log \frac{3}{4} \right]$$

$$= 0.8112$$

$$H(X_1=0/X_3=1) = - [1/1 \log 1] = 0$$

$$\text{Entropy } H(X_1=0/X_3) = \frac{4}{5} \times 0.8112 + 0$$

$$= 0.649$$

$$IG = 0.7219 - 0.649$$

$$\left( X_3 \text{ w.r.t } X_1=0 \right) = 0.0729$$

$\therefore$  Since IG of  $X_2$  is more, when  $X_1=0$ ,  $X_2$  is on left side of tree

Now consider,  $X_1=1$  (right node)

$$\text{Entropy } (X_1=1) = 0.7219$$

$$H(X_1=1/X_2=0) = - \left[ \frac{1}{4} \log \frac{1}{4} + \frac{3}{4} \log \frac{3}{4} \right]$$

$$= 0.81125$$

$$H(X_1=1 | X_2=1) = -\frac{1}{1} \log \frac{1}{1} = 0$$

$$H(X_1=1 | X_2) = \frac{4}{5} [-8/25] + \frac{1}{5} (0) = 0.649$$

$$\begin{aligned} \text{IG on } X_2 \text{ when } (X_1=1) \\ &= 0.7219 - 0.649 \\ &= 0.0729 \end{aligned}$$

Entropy of  $X_3$  when  $X_1=1$  (right side)

$$H(X_1=1) = 0.7219$$

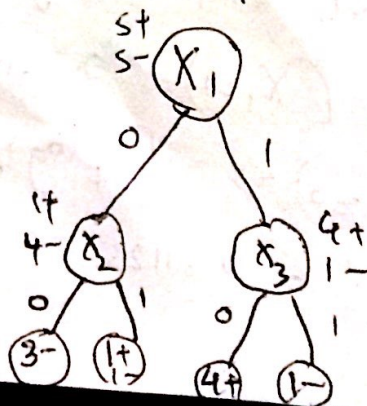
$$H(X_1=1 | X_3=0) = -\frac{4}{4} \log \frac{4}{4} = 0$$

$$H(X_1=1 | X_3=1) = -\left[\frac{1}{1} \log \frac{1}{1}\right] = 0$$

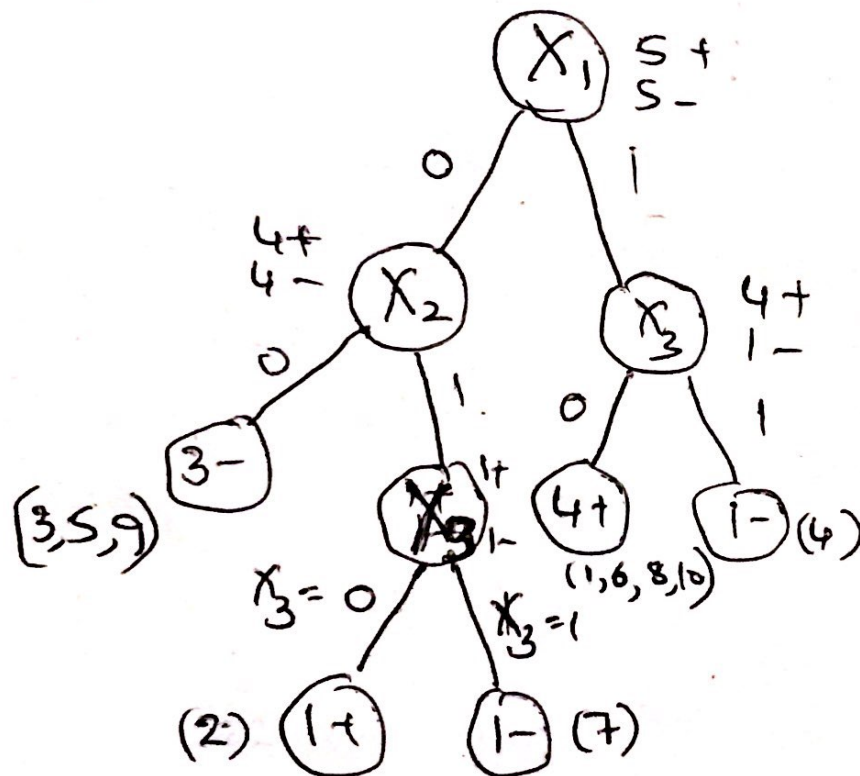
$$H(X_1=0 | X_3) = \frac{4}{5} \times 0 + \frac{1}{5} \times 0 = 0$$

$$\begin{aligned} \text{I.G on } X_3 \text{ w.r.t } X_1=1 &\Rightarrow 0.7219 - 0 \\ &= 0.7219 \end{aligned}$$

Since IG on  $X_3$  is more on right side of  $(X_1=1)$ , split it on  $X_3$ ,



From the obtained tree, we observe that the left leaf of  $X_1$  is not pure when  $X_2 = 1$ , so we will have to split it again, and since the leaf node obtained by splitting  $X_1$  on  $X_3$  is pure, no further splitting is needed.



∴ This is the final decision Tree.