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Page No.: \_\_\_\_\_

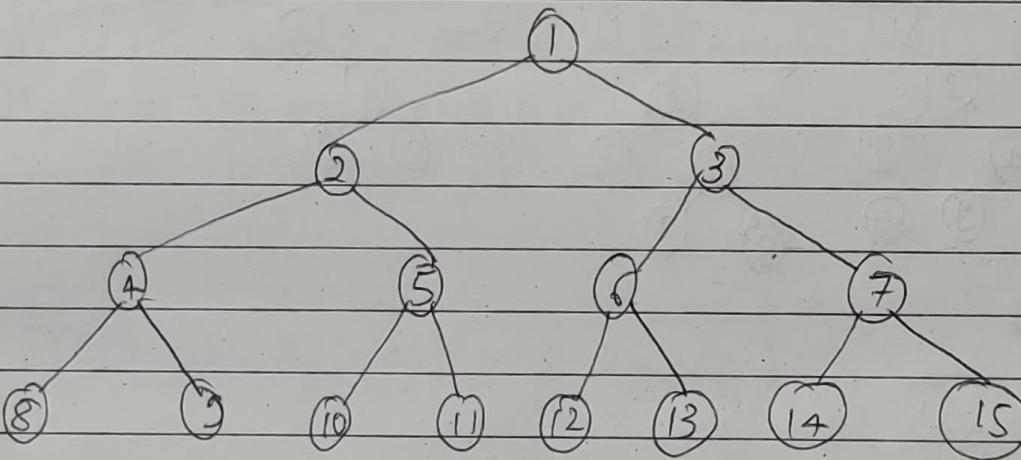
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PES2UG20CS209Assignment - I

1a) There are an infinite no of states in the state space considering all positions  $(x, y)$ . Although there is only one optimal path there are an infinite no of paths to the goal.

b) The shortest dist b/w any 2 given points is always a straight line. Therefore, the shortest path from 1 polygon vertex to any other, in the scene must consist of straight line segments joining some of the vertices of the polygons.

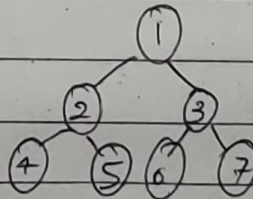
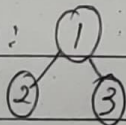
A good state space now would be all the pairs  $(x, y)$  where the pair is the vertex of an obstacle, the state space consists of all the vertices of the obstacles.

2a)



b) BFS

S0: ①, S1: ①, S2: ①

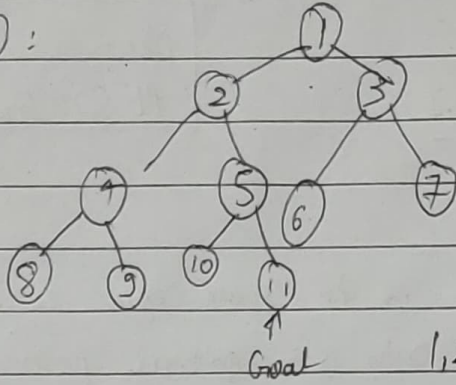




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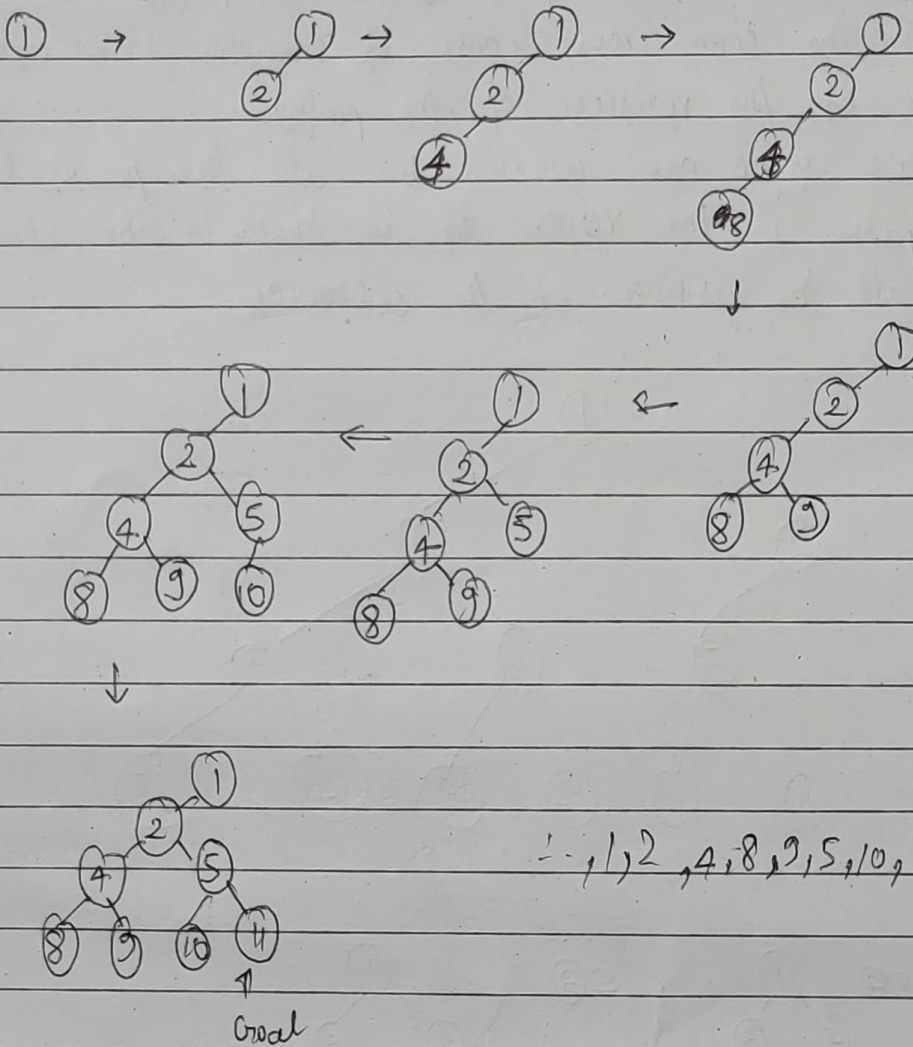
Page No.: \_\_\_\_\_

S4 :



1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11

DLS w/ limit = 3



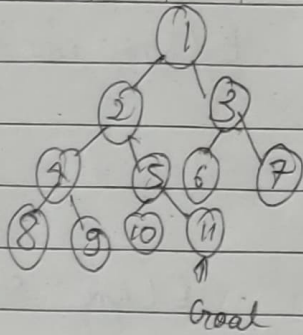




Date : \_\_\_\_\_

Page No.: \_\_\_\_\_

IDS



1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11

c)  $2k \rightarrow$  bit shift $2k+1 \rightarrow$  bit shift & + 14)  $A^*$  Search $\rightarrow$  Form of Best First Search. $F(n) = g(n) + h(n)$  is both complete & optimalA heuristic  $h(n)$  is admissible if, for every node  $n$ ,  $h(n) \leq h^*(n)$  where  $h^*(n)$  is the true cost to reach the goal state from  $n$ .

An admissible heuristic never overestimates the cost to reach the goal. It is optimistic.

If  $h(n)$  is admissible,  $A^*$  using Tree Search is optimal.

$$5) E(S) = -\frac{5}{14} \log_2 \left( \frac{5}{14} \right) - \frac{9}{14} \log_2 \left( \frac{9}{14} \right) = 0.94$$

Income	buy	comp.
high	1	3
Med	2	4
Low	2	2

$$E(\text{income} = \text{high}) = -\left[ \frac{1}{4} \log_2 \frac{1}{4} + \frac{3}{4} \log_2 \frac{3}{4} \right]$$

$$= -[-0.5 - 0.31127]$$

$$= 0.811$$

$$E(\text{income} = \text{medium}) = -\left[ \frac{2}{6} \log_2 \frac{2}{6} + \frac{4}{6} \log_2 \frac{4}{6} \right] = -[-0.520 - 0.192]$$

$$= 0.7233$$



Date: \_\_\_\_\_

Page No.: \_\_\_\_\_

$$E(\text{income} = \text{low}) = 0$$

$$E(\text{type} = \text{emp}) = -\frac{1}{7} \log \frac{1}{7} + \frac{6}{7} + \log \frac{6}{7}$$

$$E(\text{Age}) = 0$$

Types	Bugs	Comp
	P	
Employee	4	6
Student		3

$$= -[0.401 - 0.1906]$$

$$= 0.591$$

$$E(\text{type} = \text{student}) = 0.985$$

Credit Rating.

	Y	N
low	0	5
high	5	0

$$E(\text{cr} = \text{low}) = 0$$

$$E(\text{cr} = \text{high}) = -[0.4711 - 0.519]$$

$$= \boxed{0.991}$$

$$I(\text{Age}) = 0$$

$$I(\text{Income}) = \frac{4}{14} \times 0.811 + \frac{6}{14} \times 0.723 + \frac{4}{14} \times 0 = 0.2317 + 0.3098$$

$$= \boxed{0.541}$$

$$I(\text{type}) = \frac{7}{14} \times 0.591 + \frac{7}{14} \times 0.985 = 0.788$$

$$I(\text{Family Income}) = 0$$

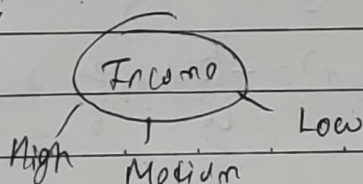
$$I(\text{cr}) = 0 + 9 \times 0.991 = 0.637$$

$$G(S, \text{income}) = 0.94 - 0.541 = 0.39$$

$$G(S, \text{type}) = 0.94 - 0.788 = 0.15$$

$$G(S, \text{CR}) = 0.94 - 0.637 = 0.30$$

Income has the highest G





Date : \_\_\_\_\_

Page No.: \_\_\_\_\_

Income = high

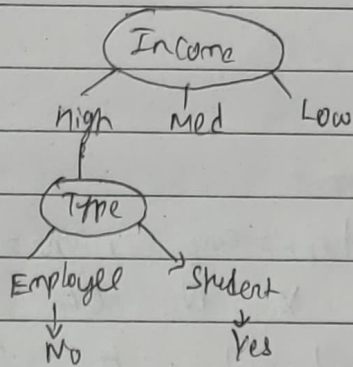
Type	CR	Buyer
Emp	L	No
Emp	L	No
Emp	H	No
Stu	H	Yes

$$E(S_{high}) = -\frac{1}{4} \log_2 \frac{1}{4} - \frac{3}{4} \log_2 \frac{3}{4} = 0.811$$

CR	Y	N
low	0	2
high	1	1

$$E() = 0$$

Type	Y	N
Stu	1	0
Emp	0	3



Income = Medium

Type	CR	Buyer
Emp	H	Y
Emp	L	N
Stu	H	Y
Stu	L	N
Emp	H	N

$$E(S_{medium}) = 0.97$$

Type	Y	N
Emp	1	2
Stu	1	1

$$E(\text{type}) = -\frac{1}{3} \log_2 \left(\frac{1}{3}\right) - \frac{2}{3} \log_2 \left(\frac{2}{3}\right)$$

$$= -[-0.528 - 0.3859]$$

$$= 0.917$$

$$I(\text{type}) = \frac{2}{5} \times 0.917 = 0.5502$$

$$G(S_{medium} \text{ type}) = 0.4198$$



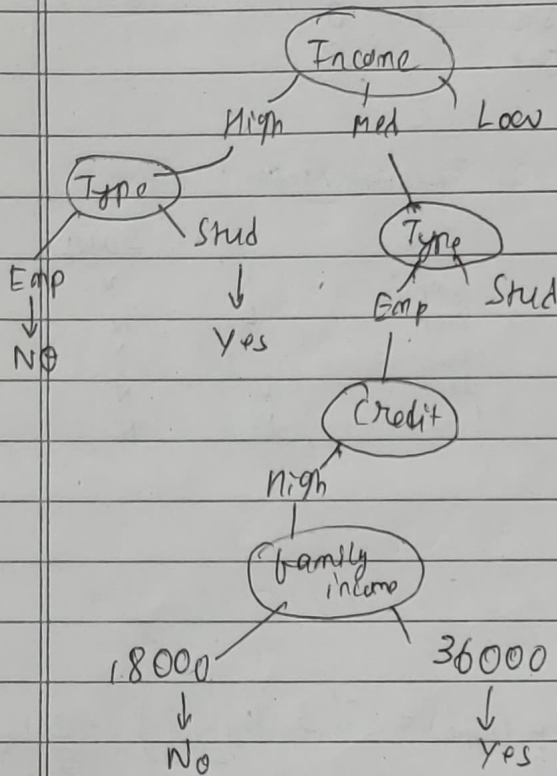
$C_n$	Y	N
high	2	1
Low	0	2

$$E(\text{high}) = -\left[\frac{2}{3} \log_2 \frac{2}{3} + \frac{1}{3} \log_2 \frac{1}{3}\right]$$

$$= -[-0.584 - 0.528] = 1.11$$

$$I(CR) = \frac{3}{5} \times 1.11 = 0.66$$

$$G(S, CR) = 0.97 - 0.66 = 0.31$$



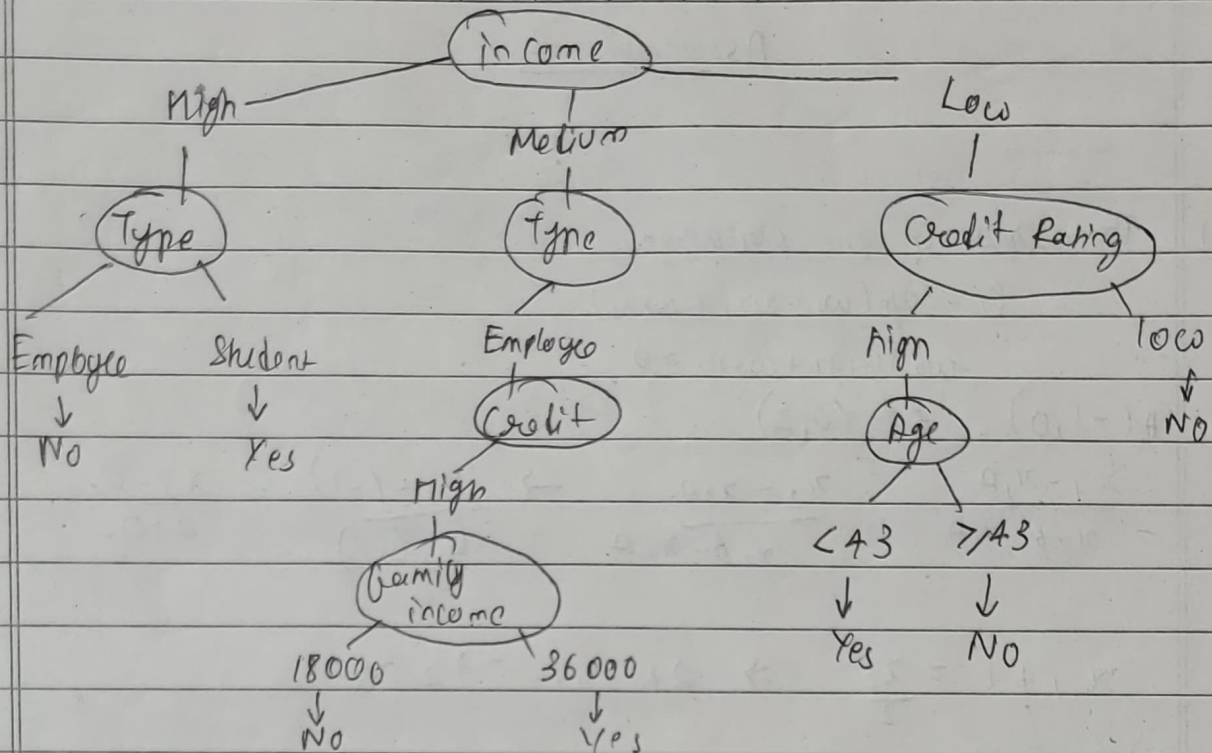
	type = emp			type = stud	
CR	Y	N	CR	Y	N
high	1	1	H	1	0
low	0	1	L	0	1

income = low

type	CR	buys
Shu	H	Y
Shu	H	N
Shu	H	Y
Shu	L	N

Attributes

age - 12    income - 3    type - 2    family - 13    CR - 2    Buys - 2



6. a) The instance space is 14  
b)  $1 + [13 \times 4 \times 3 \times 4 \times 3 \times 3] = 1957$   
c)  $14 \times 5 \times 4 \times 15 \times 4 \times 4 = 67200$   
d) The size of the concept space is 6.

5. b) rule

$(\text{Income} = \text{high} \vee \text{type} = \text{Student}) \wedge (\text{Income} = \text{medium} \vee ((\text{type} = \text{employee}) \vee \text{credit} = \text{high} \vee \text{family income} = 36000)) \wedge (\text{type} = \text{student} \vee \text{credit} = \text{high}) \wedge (\text{income} = \text{low} \vee \text{credit rating} = \text{high} \vee \text{age} < 43)$

5. c) If credit rating = Low then the person will not buy a computer.

Assignment - 2

4.1 The output of the perceptron

$$O = \text{sign}(w_0 + w_1x_1 + w_2x_2)$$

$$w_0 + w_1x_1 + w_2x_2 = 0$$

$$A(-1, 0) \quad B(0, 2)$$

$$\frac{x_1 - x_1A}{x_1B - x_1A} = \frac{x_2 - x_2A}{x_2B - x_2A} \rightarrow \frac{x_1 - (-1)}{0 - (-1)} = \frac{x_2 - 0}{2 - 0}$$

$$x_1 + 1 = \frac{x_2}{2} \Rightarrow 2 + 2x_1 - x_2 = 0$$

$$w_0 = -2, w_1 = -1, w_2 = 1$$

4.2

$$A \quad B \quad A \wedge B$$

$$-1 \quad -1 \quad -1$$

$$-1 \quad 1 \quad -1$$

$$1 \quad -1 \quad -1$$

$$1 \quad 1 \quad 1$$

$$A = B + 1$$

$$1 - A + B = 0$$

$$\text{So, } 1, -1, 1 \text{ for } A=1 \text{ \& } B=-1$$

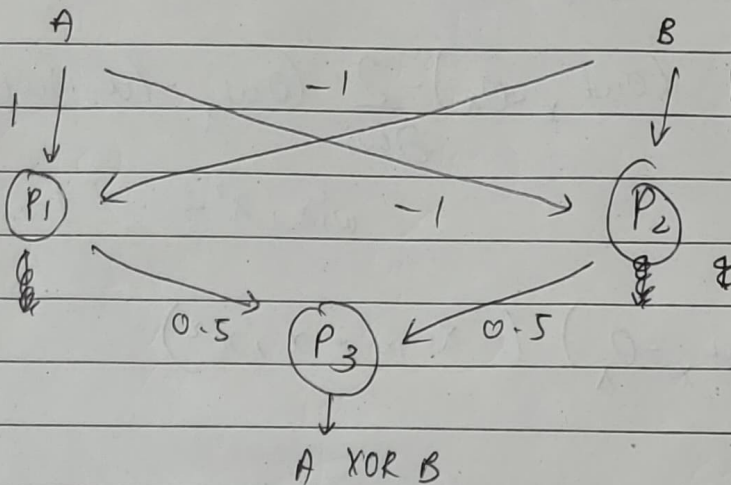
$$\Rightarrow w_0 = 1, w_1 = 1, w_2 = -1$$

$$A \text{ xor } B = (A \wedge \neg B) \vee (\neg A \wedge B)$$

$$p_1$$

$$p_2$$





4.3  $w_0 + w_1 x_1 + w_2 x_2 > 0$

A:

$w_0 = 1, w_1 = 2, w_2 = 1$

B:

$w_0 = 0, w_1 = 2, w_2 = 1$

$B(\langle x_1, x_2 \rangle) = 1$

$2x_1 + x_2 > 0$

$1 + 2x_1 + x_2 > 0$

$A(\langle x_1, x_2 \rangle) = 1$

4.5  $0 = w_0 + w_1 x_1 + w_2 x_2 + \dots + w_n x_n$

$$\frac{\partial E}{\partial w_i} = \sum_{x \in X} (o_{x_i} \neq o_{x_i}) \frac{\partial (o_{x_i} - (w_0 + w_1 x_1 + w_2 x_2 + \dots + w_n x_n))^2}{\partial w_i}$$

$$= \sum_{x \in X} (o_{x_i} - o_{x_i}) (-x_i x - x_i x^2)$$



Date: \_\_\_\_\_

Page No.: \_\_\_\_\_

Gradient descent training rule

$$\frac{\partial E_{\text{total}}}{\partial w_i} = \sum_{n \in X} (out_n - o_n) \frac{\partial}{\partial w_i} (out_n - (w_0 + w_1 x_1 + \dots + w_n x_n x^2))$$

$$= \sum_{n \in X} (out_n - o_n) (-x_1 - x_1 x^2)$$