

Assignment no 2

Name = Siddharth S. Soni

Roll no = 66

Class = BE IT

Sem = VII

Sub = IS lab

Dop	Doc	Marks	Sign

Q1) Solve the following with forward chaining or backward chaining or resolution use predicate logic as language of knowledge representation clearly specify the facts & inference rule used.

Q1) Example 1:

- 1) Every child sees some witch has both a black cat & a pointed hat.
- 2) Every witch is good or bad
- 3) Every child who sees any good witch gets candy.
- 4) Every witch that is bad has a black cat.
- 5) Every witch that is seen by any child has a pointed hat.
- 6) Prove: every child gets candy

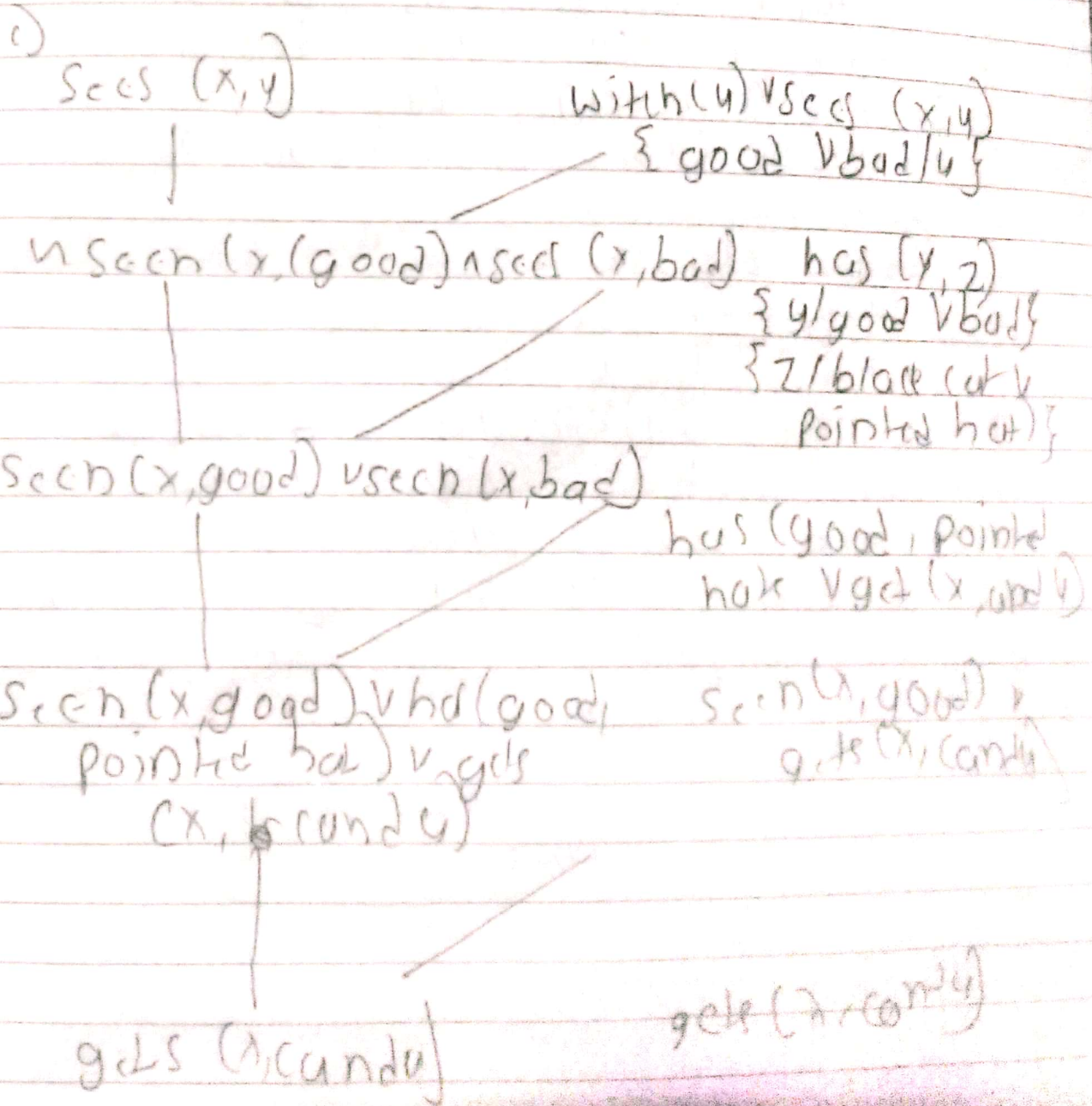
→ A) facts into fact

- 1) $\exists x \forall y (\text{child}(x), \text{witch}(y) \rightarrow \text{sees}(x, y))$
 $\neg \exists y (\text{witch}(y) \rightarrow \text{has}(y, \text{black cat}) \wedge \text{has}(y, \text{pointed hat}))$
- 2) \neg
- 2) ~~Every~~ $\exists y (\text{witch}(y) \rightarrow \text{good}(y) \vee \text{bad}(y))$
- 3) $\exists x (\text{sees}(x, y) \rightarrow (\text{witch}(y) \text{ gets candy}))$
- 4) $\forall y (\text{witch}(y) \rightarrow \text{bad}(y) \rightarrow \text{has}(y, \text{black hat}))$
- 5) $\forall y (\text{sees}(x, y) \rightarrow \text{has}(y, \text{pointed hat}))$

B) fact into CNF

- 1) $\exists x \forall y (\text{child}(x), \text{witch}(y) \rightarrow \text{sees}(x, y))$
 $\rightarrow \neg \exists y, (\text{witch}(y) \rightarrow \text{has}(y, \text{black hat}))$
 $\rightarrow \neg \exists y (\text{witch}(y) \rightarrow \text{has}(y, \text{pointed hat}))$

- 2) $\forall y (\text{Witch}(y) \rightarrow \text{good}(y))$
 $\forall y (\text{Witch}(y) \rightarrow \text{bad}(y))$
- 3) $\exists x (\text{Sees}(x, y) \rightarrow \text{Witch}(y) \rightarrow \text{good}(y))$
 $\text{get}(x, \text{and } y)$
 $\rightarrow \exists x [\text{Sees}(x, \text{good}(y) \rightarrow \text{get}(x, \text{and } y))]$
- 4) $\forall y [\text{bad}(y) \rightarrow \text{has}(y, \text{black hats})]$
- 5) $\exists y [\text{Sees}(x, y) \rightarrow \text{hd}(y, \text{pointed hat})]$
 $\rightarrow \sim \forall y [\text{Sees}(x, y) \rightarrow \text{hd}(y, \text{black hat})]$



2) Example 2:

1) Every-boy or girl is a child

ans)

1) $\forall x (\text{boy}(x) \vee \text{girl}(x) \rightarrow \text{child}(x))$

2) $\forall (\text{child}(y) \rightarrow \text{gets}(y, \text{doll}) \vee \text{gets}(y, \text{train}) \vee \text{gets}(y, \text{coal}))$

3) $\exists w (\text{boy}(w) \rightarrow \neg \text{gets}(w, \text{doll}))$

4) for all $z (\text{child}(z) \wedge \text{bad}(z) \rightarrow \text{gets}(z, \text{coal}))$
 $\forall y (\text{child}(y) \rightarrow \neg \text{gets}(y, \text{train}))$

5) $\text{child}(\text{sam}) \rightarrow \text{gets}(\text{sam}, \text{coal})$

to prove $\neg \text{child}(\text{sam}) \rightarrow \text{bad}(\text{sam})$

(NFC clause)

1) $\neg \text{boy}(x) \vee \neg \text{child}(x)$

2) $\neg \text{girl}(x) \vee \neg \text{child}(x)$

3) $\neg \text{child}(y) \vee \neg \text{gets}(y, \text{doll}) \vee \neg \text{gets}(y, \text{train}) \vee \neg \text{gets}(y, \text{coal})$

4) $\neg \text{boy}(w) \vee \neg \neg \text{gets}(w, \text{doll})$

5) $\neg \text{child}(z) \vee \neg \text{bad}(z) \vee \neg \text{gets}(z, \text{coal})$

6) $\neg \text{child}(\text{sam}) \rightarrow \text{gets}(\text{sam}, \text{coal})$

7) $\text{bad}(\text{sam})$

Resolution:

4) 1 child (2) or 1 bad (2) or get (2, coal)

6) bad (ram)

7) 1 child (ram) or gets (ram, coal)

Substituting 2 by ram

1) 1a) 1 boy (x) or child (x)

boy (ram)

2) child ram (substituting x by ram)

7) 1 child (ram) or gets (ram, coal)

8) child (ram)

9) gets (ram, coal)

2) 1 child (y) or gets (y, doll) or get

5) child (ram)

9) gets (ram, coal)

2) gets (ram, doll) or gets (ram, coal)

3) 1 boy (w) or 1 gets (w, doll)

5) boy (ram)

12) 1 get (ram, doll) (substituting w by ram)

11) gets (ram, doll) or get (ram, train)

12) 1 gets (ram, doll)

13) gets (ram, coal)

6) 1a) get (ram, coal)

13) gets (ram, coal)

none 1 bad (ram) 8 proved

82) Difference between STRIPS and AOL

STRIPS language	AOL
1) only allow positive literals in the state for eg. A valid sentence in STRIPS is expressed as	1) can support both positive & negative literals for eg :- some sentence is expressed as stupid & ugly
2) STRIPS stand for Stanford Research Institute problem solver.	2) stands for Action description language
3) makes use of closed world assumption (i.e.) unmentioned literal is false	3) make use of open world assumption (i.e.) unmentioned literal is unknown
4) goals are conjunctions for ex :- (Intelligent & Beautiful)	4) Goals may involve conjunctions & disjunctions eg. (Intelligent \wedge Beautiful) OR
5) Does not support equality	5) Equality predicate ($x=y$) is permitted
6) Does not have support for types	6) Support for types for each variable is provided

P4) $P(B)$
0.001

Burglary

Earthquake

Alarm

John
calls

Mary
calls

B	E	$P(A)$
F	T	0.95
T	F	0.94
F	T	0.29
F	F	0.001

A	$P(J)$
T	0.09
F	0.05

M	$P(M)$
T	0.70
F	0.01

1) The topology of the network indicates that
- Burglary and earthquake affect
probability of alarm going off

- Whether John and Mary call depends only
on alarm

2) Mary listening to loud music
John confusing phone ringing to sound
alarm can be read from network
only implicitly or uncertainty
associated by calling out 'huh'

3) The condition probability table in
n/w gives probability for value of
random variable dependent
on combination of values of
parent nodes

4) A variable with no parent has only one row, representing prior probabilities of each possible value of the variable.

5) Every entry in full joint probability distribution can be calculated from information in Bayesian network.

6) A generic entry in joint distribution probability of conjunction of particular assignment to each variable $P(x_1 = x_1, \dots, x_n = x_n)$ is divided as $P(x_1, \dots, x_n)$.

7) Value of this entry is $P(x_1, \dots, x_n)$
 $= P(i, \text{parents}(x_i))$, where $\text{parents}(x_i)$ denotes the specific values of the variables $\text{parents}(x_i)$
 $= P(i, a) P(m, a) P(a, b, m, c) P(c, d, e, m)$
 $= 0.09 \times 0.07 \times 0.001 \times 0.999 \times 0.998$
 $= 0.000629$

12) Bayesian network

