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

Q1. Example 1:

- ① Every child sees some witch No 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 FOL.

- 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 $\exists y (\text{witch}(y) \rightarrow \text{good}(y) \vee \text{bad}(y))$
- ③ $\exists x ((\text{sees}(x, y) \rightarrow (\text{witch}(y) \rightarrow \text{good}(y))) \rightarrow \text{get}(x, \text{candy}))$
- ④ $\exists y ((\text{witch}(y) \rightarrow \text{bad}(y)) \rightarrow \text{has}(y, \text{black hat}))$
- ⑤ $\exists y (\text{sees}(x, y) \rightarrow \text{has}(y, \text{pointed hat}))$

(B) FOL 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) $\exists x [\text{sees}(x, y) \rightarrow \text{witch}(y) \rightarrow \text{good}(y)] \rightarrow \text{gets}(x, \text{candy})$

$\rightarrow \exists x [\text{see}(x, y) \rightarrow \text{witch}(y) \rightarrow \text{good}(y)]$

$\rightarrow \exists x [\text{sees}(x, \text{good}(y)) \rightarrow \text{gets}(x, \text{candy})]$

(4) $\exists y [\text{bad}(y) \rightarrow \text{has}(y, \text{black hats})]$

(5) $\exists y [\text{seen}(x, y) \rightarrow \text{has}(y, \text{black hat})]$

c)

$\text{sees}(x, y)$

$\text{witch}(y) \vee \text{seen}(x, y)$

$\{\text{good} \vee \text{bad} / y\}$

$\neg \text{seen}(x, \text{good}) \wedge \text{seen}(x, \text{bad})$

$\text{has}(y, z)$

$\{y / \text{good} \vee \text{bad}\}$

$\{z / \text{black cat} \vee$

$\text{pointed hat}\}$

$\text{seen}(x, \text{good}) \vee \text{seen}(x, \text{bad})$

$\text{has}(\text{good}, \text{pointed}$

$\text{hats} \vee \text{get}(x, \text{candy})$

$\text{seen}(x, \text{good}) \vee \text{has}(\text{good},$
 $\text{pointed hat}) \vee \text{gets}$

(x, candy)

$\text{seen}(x, \text{good}) \vee$

$\text{gets}(x, \text{candy})$

$\text{gets}(x, \text{candy})$

$\text{gets}(x, \text{candy})$

(2) Example 2:

- ① Every boy or girl is a child
- ② Every child gets a doll or a train or a lump of a coal.
- ③ No boy gets any doll
- ④ Every child who is bad gets any lump of coal
- ⑤ No child gets a train.
- ⑥ Ram gets lump of coal.
- ⑦ Prove Ram is bad.

- ① $\forall x (\text{boy}(x) \text{ or } \text{girl}(x) \rightarrow \text{child}(x))$
- ② $\forall y (\text{child}(y) \rightarrow \text{gets}(y, \text{doll}) \text{ or } \text{gets}(y, \text{train}) \text{ or } \text{gets}(y, \text{coal}))$
- ③ $\forall w (\text{boy}(w) \rightarrow \text{!gets}(w, \text{doll}))$
- ④ for all $z = (\text{child}(z) \text{ and } \text{bad}(z)) \rightarrow \text{gets}(z, \text{coal})$
 $\forall y \text{ child}(y) \rightarrow \text{!gets}(y, \text{train})$
- ⑤ $\text{child}(\text{ram}) \rightarrow \text{gets}(\text{ram}, \text{coal})$
 To prove $\text{child}(\text{ram}) \rightarrow \text{bad}(\text{ram})$

(CNF clauses)

- ① $\text{!boy}(x) \text{ or } \text{child}(x)$
 $\text{!girl}(x) \text{ or } \text{child}(x)$
- ② $\text{!child}(y) \text{ or } \text{gets}(y, \text{doll}) \text{ or }$
 $\text{gets}(y, \text{train}) \text{ or } \text{gets}(y, \text{coal})$
- ③ $\text{!boy}(w) \text{ or } \text{!gets}(w, \text{doll})$
- ④ $\text{!child}(z) \text{ or } \text{!bad}(z) \text{ or } \text{gets}(z, \text{coal})$
- ⑤ $\text{!child}(\text{ram}) \rightarrow \text{gets}(\text{ram}, \text{coal})$
- ⑥ $\text{bad}(\text{ram})$

Resolution:

- (4) ! child (z) or ! bad (z) or get (z, coal)
 - (5) bad (ram)
 - (6) ! child (ram) or gets (ram, coal)
Substituting z by ram
 - (1) (a) ! boy (x) or child (x)
boy (ram)
 - (2) child ram (substituting x by ram)
 - (7) ! child (ram) or gets (ram, coal)
 - (8) child (ram)
 - (9) gets (ram, coal)
 - (2) ! child (y) (or gets (y, doll) or gets (y, train) or
gets (y, coal))
 - (8) child (ram)
 - (10) gets (ram, doll) or gets (ram, train) or gets
(ram, coal) (substituting y by ram)
 - (9) gets (ram, doll)
 - (10) gets (ram, doll) or gets (ram, train) or gets
(ram, coal)
 - (11) gets (ram, doll) or gets (ram, train) or coal
 - (3) ! boy (w) or ! gets (w, doll)
 - (5) boy (ram)
 - (12) ! get (ram, doll) (substituting w by ram)
 - (11) gets (ram, doll) or gets (ram, train).
 - (12) ! gets (ram, doll)
 - (13) gets (ram, coal)
 - (6) (a) get (ram, coal)
 - (16) gets (ram, coal)
- Hence, bad (ram) is proved.

Q.2 Differentiate between STRIPS and ADL

STRIPS language

ADL

① Only allow positive literals in the states
for eg: A valid sentence is STRIPS is expressed as \Rightarrow Intelligent A Beautiful

① can support both positive & negative literals
for eg: - same sentence is expressed as \Rightarrow stupid A - ugly

② STRIPS stand For standard Research Institute Problem solver

② stands for Action Description Language

③ Makes use of closed world assumption (i.e) unmentioned literals are False.

③ makes use of open world Assumption (i.e) unmentioned literals are unknown.

④ we only can find ground literals in goals .
For eg: - Intelligent A Beautiful

④ we can find qualified variables in goal for eg: $Fx A1(p_1, x) \wedge A1(p_2, x)$ is the goal of having p_1 & p_2 in the same place in the eg of blocks

⑤ Goals are conjunction for eg: (Intelligent n Beautiful)

⑤ Goals may involve conjunctions & disjunction For eg: - (Intelligent n (Beautiful \wedge Rich))

⑥ Effects are conjunctions

⑥ conditional effects are allowed : when P: E means E is an effect only if P is satisfied -

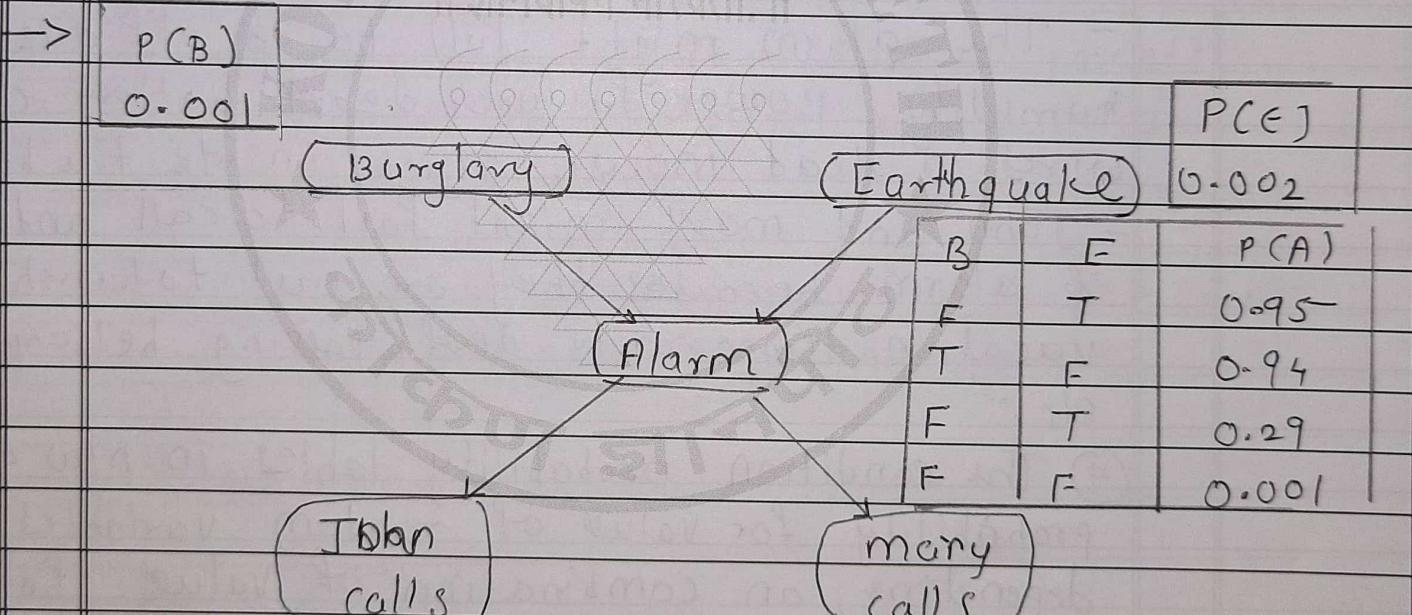
⑦ Does not support equality

⑦ Equality predicate ($x = y$)
is built in

⑧ Does not have support for types

⑧ Support - For types for -
eg : The variable
p : person

Q.4. You have two neighbors J and M, who have promised to call you at work when there is an alarm. J always calls when he hears the alarm, but sometimes confuses telephone ringing with alarms & calls them too. It likes loud music and sometimes misses the alarm together. Given the evidence of who has or has not called we would like to estimate the probability of burglary. Draw a Bayesian network for this domain with suitable probability table.



A	$P(A)$
T	0.09
F	0.05

A	$P(M)$
T	0.70
F	0.01

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

- Whether John and Mary call depends only on alarm.
 - They do not perceive any burglaries directly they do not notice minor earthquakes and they do not confer before calling.
- (2) Many listening to loud music & John confusing phone ringing to sound of alarm can be read from network only implicitly as uncertainty associated to calling at work.
- (3) The probability actually summarize potentially infinite sets of circumstance
- The alarm might fail to go off due to high humidity, power failure, dead battery, cut wires, a dead mouse stuck inside the bell, etc.
 - John and Mary might fail to call and report the alarm because they are out to launch on vacation, temporarily deaf, passing helicopter etc.
- (4) The condition probability tables in NW gives the probability for values of random variables depending on combination of values for the parent nodes.
- (5) Each row must be sum to 1, because entries represent exhaustive set of causes for variable.
- (6) All variables are Boolean
- (7) In general, a table for a Boolean variable with k parents contains 2^k independently specific probabilities.

- (8) A variable with no parents has only one row, representing prior probabilities of each possible value of the variable.
- (9) Every entry in full joint probability distribution can be calculated from information in Bayesian network.
- (10) A generic entry in joint distribution is probability of a conjunction of particular assignments to each variable $P(x_1 = x_1, \dots, x_n = x_n)$ abbreviated as $P(x_1, \dots, x_n)$
- (11) The value of this entry is $P(x_1, \dots, x_n) = \prod_{i=1}^n p(x_i | \text{parents}(x_i))$, where $\text{parents}(x_i)$ denotes the specific values of the variable parents (x_i)

$$\begin{aligned}
 &= P(j \wedge m \wedge a \wedge \sim b \wedge \sim e) \\
 &= P(j|a) P(m|a) P(a \wedge b \wedge \sim e) P(\sim b) P(\sim e) \\
 &= 0.09 \times 0.07 \times 0.001 \times 0.999 \times 0.998 \\
 &= 0.000628
 \end{aligned}$$

(12) Bayesian Network

