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Q1.

Solve the following with forward chaining or backward chaining or resolution w.e predicate logic as language of knowledge representation clearly specify the facts and inference rule used.

Example 1 :

1) Every child sees some witch who 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) $\exists x \forall y (child(x), witch(y) \rightarrow sees(x, y))$

$\sim \exists y (witch(y) \rightarrow has(y, black\ cat) \wedge has(y, pointed\ hat))$

2) $\exists y (witch(y) \rightarrow good(y) \vee bad(y))$

3) $\exists x ((sees(x, y) \rightarrow (witch(y) \vee bad(y))) \rightarrow get(x, candy))$

4) $\forall y ((witch(y) \rightarrow bad(y)) \rightarrow has(y, black\ hat))$

5) $\forall y (sees(x, y) \rightarrow has(y, pointed\ hat))$

B) FOL into CNF

1) $\exists x \forall y (child(x), witch(y) \rightarrow sees(x, y))$

2) $\rightarrow \sim \exists y, (witch(y) \rightarrow has(y, black\ cat))$

$\rightarrow \sim \exists y (witch(y) \rightarrow has(y, pointed\ hat))$

3) $\forall y (witch(y) \rightarrow good(y))$

$\forall y (witch(y) \rightarrow bad(y))$

4) $\exists x ((sees(x, y) \rightarrow witch(y) \rightarrow good(y)) \rightarrow gets(x, candy))$

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

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

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

→ $\sim \forall y [\text{seen}(x, y) \rightarrow \text{has}(y, \text{black hat})]$

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

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

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

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

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

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

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

$\text{pointed hat} \}$

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

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

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

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

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

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

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

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

Example 2:

1) Every boy or girl is a child.

2) Every child gets a doll or a train or a lump of 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) $\forall w (\text{boy}(w) \rightarrow \neg \text{gets}(w, \text{doll}))$

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

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) $\neg \text{boy}(x) \text{ or } \text{child}(x)$

$\neg \text{girl}(x) \text{ or } \text{child}(x)$

2) $\neg \text{child}(y) \text{ or } \text{gets}(y, \text{doll}) \text{ or } \text{gets}(y, \text{train}) \text{ or } \text{gets}(y, \text{coal})$

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

4) $\neg \text{child}(z) \text{ or } \neg \text{bad}(z) \text{ or } \text{gets}(z, \text{coal})$

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

6) $\text{bad}(\text{ram})$

Resolution

4) $\neg \text{child}(z) \text{ or } \neg \text{bad}(z) \text{ or } \text{get}(z, \text{goal})$

6) $\text{bad}(\text{ram})$

FOR EDUCATIONAL USE

7) ! child (from) or gets (ram, coal)

substituting 2 by ram

17 (a) ! boy (x) or child (x) boy ram

8) child ram (substituting x by ram)

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

8) child (ram)

9) gets (ram, coal)

10) ! child (y) (or gets (y, doll) or gets (y, train) or gets (y, coal))

7) child (ram)

10) gets (ram, doll) or gets (ram, train) or gets (ram, coal)

9) gets (ram, coal)

10) gets (ram, doll) or gets (ram, train) or gets (ram, coal)

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

12) gets (ram, coal)

6) <a> get (ram, coal)

13) gets (ram, coal)

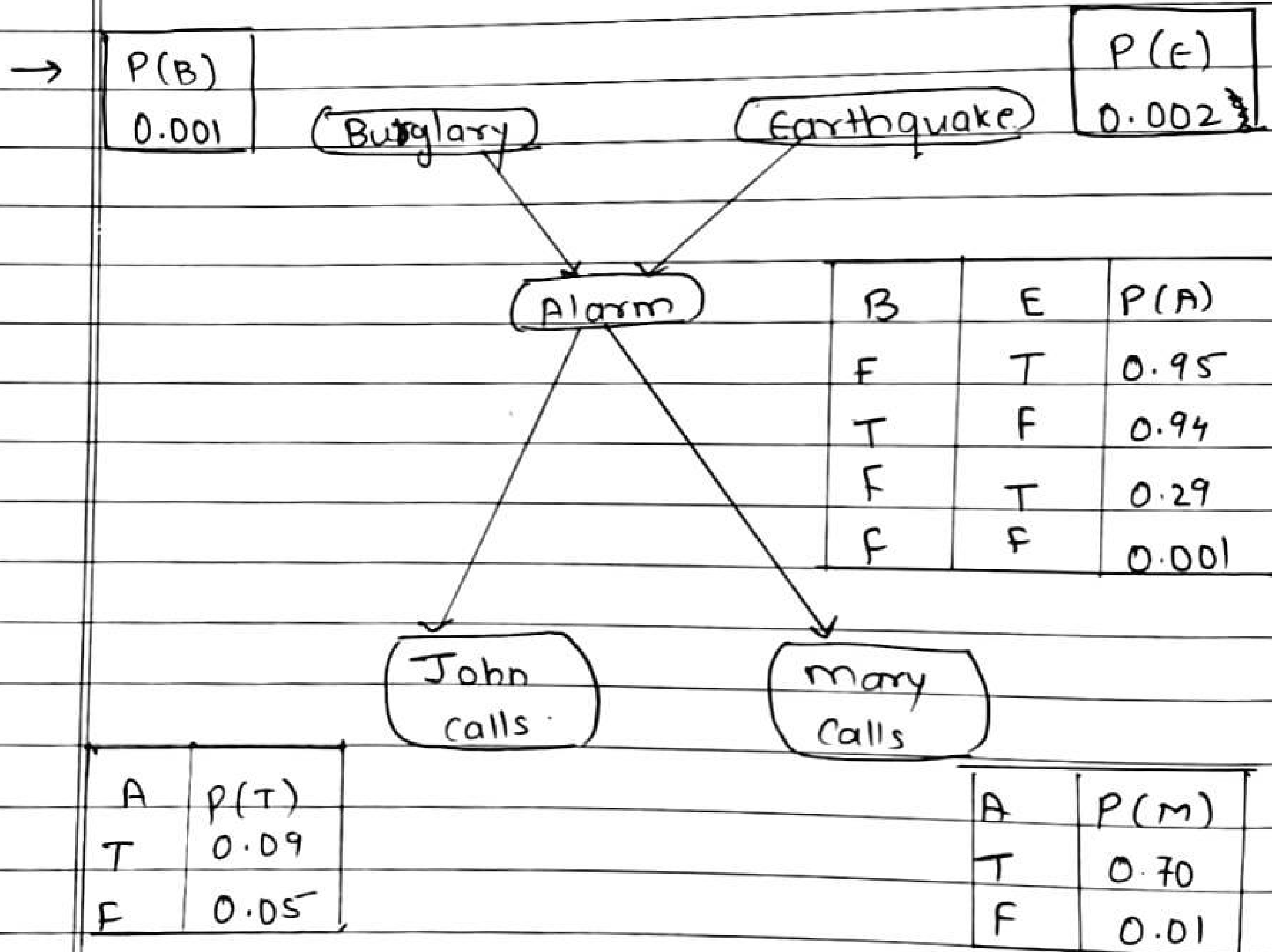
Hence, bad (ram) is proved.

Q.2

Differentiate between STRIPS and ADL

Strips Language	ADL
<p>① Only allow positive literals in the states. for e.g. : A Valid sentence in STRIPS is expressed as $\Rightarrow \text{Intelligent} \wedge \text{Beautiful}$</p>	<p>① Can support both positive & negative literals for e.g. :- Same sentence is expressed as $\Rightarrow \text{Stupid} \wedge \neg \text{ugly}$</p>
<p>② STRIPS stands for standard Research Institute Problem Solver</p>	<p>② Stands for Actions Description Language</p>
<p>③ Makes use of closed world assumption (i.e.) unmentioned literals are false.</p>	<p>③ Makes use of open world Assumption (i.e.) unmentioned literals are unknown.</p>
<p>④ We only can find ground literals in goals for e.g. :- Intelligent \wedge Beautiful</p>	<p>④ We can find qualified variables in goal for e.g. :- $\exists x \text{ At}(P_1, x) \wedge \text{At}(P_2, x)$ is the goal of having P_1 & P_2 in the same place in e.g. of blocks.</p>
<p>⑤ Goals are conjunctions for e.g. :- (Intelligent \wedge beautiful)</p>	<p>⑤ Goals may involve conjunction & disjunctions for e.g. :- (Intelligent \wedge (Beautiful \vee Rich))</p>

Q4) You have two neighbors J and M, who have promised to call you at work when they hear the alarm. J always calls when he hears the alarm, but sometimes confused telephone ringing with alarms & calls then too. M 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.



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) Mary 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 circumstances.

- 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 of alarm because they are out to lunch, on vacation, temporarily deaf, passing helicopter, etc.

4) The condition probability tables in nlw gives 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 cases 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 $P(x_1 = x_1 \wedge \dots \wedge x_n = x_n)$ 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 variables $\text{parents}(x_i)$

$$= P(j|a) P(m|a) P(a|b,n,e)$$

$$= P(j|a) P(m|a) P(a|b,n,e) P(b) P(e)$$

$$= 0.09 \times 0.07 \times 0.001 \times 0.999 \times 0.998$$

$$= 0.000628$$

12) Bayesian Network.

