CS267A: Homework #4

Tianyu Zhang

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Problem 1

Solution:

What is the probability that there is a burglary or earthquake given both John and Mary call: 0.092088197

- 1. What is the probability that there is an earthquake given no burglary and Mary calls: 0.14564687
- 2. Please find the following my code that generates the solution.

```
Myprogram.pl:
```

```
t (0.001) :: burglary.
t (0.002) :: earthquake.
t(0.95)::p_alarm1.
t (0.94) :: p_alarm2.
t(0.29)::p_alarm3.
t (0.001) :: p_alarm4.
alarm :- burglary, earthquake, p_alarm1.
alarm :- burglary, \+earthquake, p_alarm2.
alarm :- \+burglary, earthquake, p_alarm3.
alarm :- \+burglary, \+earthquake, p_alarm4.
t(0.9) :: call_j1.
t(0.05) :: call_j2.
call_john :- alarm, call_j1.
call_john :- \+alarm, call_j2.
t(0.7) :: call_m1.
t (0.01) :: call_m2.
call_mary := alarm, call_m1.
call_mary :- \+alarm, call_m2.
Myevidence.pl:
%%% The data:
```

```
evidence (burglary, false).
evidence (earthquake, false).
evidence (call_john, false).
evidence (call_mary, false).
evidence (burglary, true).
evidence (earthquake, false).
evidence (call_john, true).
evidence (call_mary, false).
evidence (burglary, false).
evidence (earthquake, false).
evidence (call_john, false).
evidence (call_mary, false).
evidence (burglary, false).
evidence (earthquake, false).
evidence (call_john, false).
evidence (call_mary, true).
evidence (burglary, true).
evidence (earthquake, false).
evidence (call_john, true).
evidence (call_mary, true).
evidence (burglary, true).
evidence (earthquake, true).
evidence (call_john, true).
evidence (call_mary, true).
evidence (burglary, false).
evidence (earthquake, false).
evidence (call_john, false).
evidence (call_mary, false).
evidence (burglary, false).
evidence (earthquake, false).
evidence (call_john, false).
evidence (call_mary, true).
question1.pl:
% Your model here
0.375:: burglary.
0.125:: earthquake.
1::p_alarm1.
1::p_alarm2.
0.29:: p_alarm3.
0::p_alarm4.
alarm :- burglary, earthquake, p_alarm1.
alarm :- burglary, \+earthquake, p_alarm2.
alarm :- \+burglary, earthquake, p_alarm3.
```

```
alarm :- \+burglary, \+earthquake, p_alarm4.
1:: call_j1.
0:: call_{-i}2.
call_john :- alarm, call_j1.
call_john :- \+alarm, call_j2.
0.66666667:: call_m1.
0.4:: call_m2.
call_mary :- alarm, call_m1.
call_mary :- \+alarm, call_m2.
%run one at a time
evidence (burglary, false).
evidence (call_mary, true).
query (earthquake).
%run one at a time
out1 :- burglary.
out1 :- earthquake.
evidence (call_john, false).
evidence (call_mary, true).
query (out1).
```

Problem 2

```
Solution: There is one in T means, 1 minus the possibility that no one is in T: \Pr(\exists x.T(x)) = 1 - \Pr(\omega_8) = 1 - 0.06 = 0.94.
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```
Solution: Add up all the scenarios of A,B,C appearing in the case: \Pr(A) = \Pr(\omega_1) + \Pr(\omega_2) + \Pr(\omega_4) + \Pr(\omega_5) = 0.16 + 0.16 + 0.04 + 0.04 = 0.40. \Pr(B) = \Pr(\omega_1) + \Pr(\omega_2) + \Pr(\omega_3) + \Pr(\omega_6) = 0.16 + 0.16 + 0.24 + 0.24 = 0.80. \Pr(C) = \Pr(\omega_1) + \Pr(\omega_3) + \Pr(\omega_4) + \Pr(\omega_7) = 0.16 + 0.24 + 0.04 + 0.06 = 0.50.
```

Problem 3

```
Solution:

(a) \Pr(T(\text{Alice,Pixar})) = \Pr(\omega_1) + \Pr(\omega_2) = 3/8.

(b) \Pr(\exists x. \ T(\text{Alice,}x)) = \Pr(\omega_1) + \Pr(\omega_2) + \Pr(\omega_3) = 3/4.

(c) \Pr(\exists x, y. \ T(x, y)) = \Pr(\omega_1) + \Pr(\omega_2) + \Pr(\omega_3) = 3/4.

(d) \Pr(x. \ T(x,\text{Brown})) = \Pr(T(\text{Alice,Brown}) \land T(\text{Carol,Brown})) = 0
```

```
Solution: this probabilistic database is not tuple-independent. For tuple-independence database:  \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar}) \land \mathsf{T}(\mathsf{Carol},\mathsf{UPenn})) = \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar})) \times \Pr(\mathsf{T}(\mathsf{Carol},\mathsf{UPenn})). \\ \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar}) \land \mathsf{T}(\mathsf{Carol},\mathsf{INRIA})) = \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar})) \times \Pr(\mathsf{T}(\mathsf{Carol},\mathsf{INRIA})). \\ \text{However, } \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar})) = \Pr(\omega_1) + \Pr(\omega_2) = 1/8 + 1/4 = 3/8, \\ \Pr(\mathsf{T}(\mathsf{Carol},\mathsf{UPenn})) = \Pr(\omega_1) = 1/8, \\ \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar})) \times \Pr(\mathsf{T}(\mathsf{Carol},\mathsf{UPenn})) = (3/8)(1/8) = 3/64 \\ \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar})) \wedge \Pr(\mathsf{T}(\mathsf{Carol},\mathsf{UPenn})) = 1/8 \\ \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar})) \times \Pr(\mathsf{T}(\mathsf{Carol},\mathsf{INRIA})) = (3/8)(1/4) = 3/32 \\ \Pr(\mathsf{T}(\mathsf{Alice},\mathsf{Pixar}) \wedge \mathsf{T}(\mathsf{Carol},\mathsf{INRIA})) = 1/4 \\ \mathsf{They are not equal. So it is not tuple-independent.}
```

Problem 4

1.

```
Solution: Let A = \text{Alice}, B = \text{Bob}, C = \text{Charlie}. The set of all possible world databases of T_H is \{\{A,B,C\},\{A,B\},\{A,C\},\{B,C\},\{A\},\{B\},\{C\},\{\}\}\}. The probabilities of each world are: \Pr(\{A,B,C\}) = \Pr(A,B,C) = 0.7 \times 0.4 \times 0.9 = 0.252. \Pr(\{A,B\}) = \Pr(A,B,C) = 0.7 \times 0.4 \times 0.1 = 0.028. \Pr(\{A,C\}) = \Pr(A,B,C) = 0.7 \times 0.6 \times 0.9 = 0.378. \Pr(\{B,C\}) = \Pr(\neg A,B,C) = 0.3 \times 0.4 \times 0.9 = 0.108. \Pr(\{A\}) = \Pr(A,B,C) = 0.7 \times 0.6 \times 0.1 = 0.042. \Pr(\{A\}) = \Pr(\neg A,B,\neg C) = 0.3 \times 0.4 \times 0.1 = 0.012. \Pr(\{C\}) = \Pr(\neg A,B,C) = 0.3 \times 0.6 \times 0.9 = 0.162. \Pr(\{C\}) = \Pr(\neg A,B,C) = 0.3 \times 0.6 \times 0.1 = 0.018.
```

```
Solution: 1. Pr(\exists x.H(x))
             = 1 - \Pr(\forall x. \neg H(x))
             = 1 - (1 - H(A))(1 - H(B))(1 - H(C))
             = 1 - (1 - 0.7)(1 - 0.4)(1 - 0.9)
             = 1 - 0.3 \times 0.6 \times 0.1
             = 1 - 0.018 = 0.982
             2. Pr(\exists x. H(x) \land E(x, AI))
             = 1 - \Pr(\forall x. \neg H(x) \lor \neg E(x, AI))
             = 1 - \prod_{i} Pr(\neg H(i) \vee \neg E(i, AI))
             = 1 - \prod_{i} (1 - Pr (H(i) \wedge E(i, AI)))
             = 1 - \prod_{i} (1 - Pr(H(i)) \times Pr(E(i, AI)))
             = 1 - (1 - Pr(H(A)) Pr(E(A,AI))) (1 - Pr(H(B)) Pr(E(B,AI))) (1 - Pr(H(C)) Pr(E(C,AI)))
             = 1 - (1 - 0.7 \times 0.7)(1 - 0.4 \times 0.3)(1 - 0.9 \times 0)
             = 1 - (1 - 0.49)(1 - 0.12) = 1 - 0.51 \times 0.88 = 1 - 0.45
             = 0.55.
             3. \Pr(\exists x \exists y. H(x) \land E(x,y))
             = 1 - Pr(\forall x \forall y. \neg H(x) \lor \neg E(x, y))
             = 1 - \prod_{i} (\Pr(\forall y. \neg H(i) \lor \neg E(i, y)))
             = 1 - \prod_{i} (\Pr(\neg H(i) \lor \forall y. \neg E(i, y)))
             = 1 - \prod_{i} (1 - \Pr(H(i) \wedge \exists y. E(i, y)))
             = 1 - \prod_{i} (1 - \Pr(H(i)) \times \Pr(\exists y. E(i, y)))
             = 1 - \prod_{i} (1 - \Pr(H(i)) X (1 - \Pr(\forall y. \neg E(i, y))))
             = 1 - \prod_{i} (1 - \Pr(H(i)) \times (1 - \prod_{i} (1 - \Pr(E(i,j)))))
             = 1 - (1 - Pr(H(A)) \times (1 - \prod_{i} (1 - Pr(E(A,j)))))(1 - Pr(H(B)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - \prod_{i} (1 - Pr(E(B,j)))))(1 - Pr(H(C)) \times (1 - Pr(E(B,j))))(1 - Pr(E(B,j)))(1 - Pr(E(B,j))(1 - Pr(E(B,j)))(1 - Pr(E(B,j)))(1 - Pr(E(B,j)))(1 - Pr(E(B,j))(1 - Pr(E(B,j)))(1 - Pr(E(B,j))(1 - Pr
             Pr(E(C,j))))
             = 1 - (1 - Pr(H(A)) \times (1 - (1 - Pr(E(A,AI)))(1 - Pr(E(A,PL)))))(1 - Pr(H(B)))
             X (1 - (1 - Pr(E(B,AI)))(1 - Pr(E(B,PL)))))(1 - Pr(H(C))
             X (1 - (1 - Pr(E(C,AI)))(1 - Pr(E(C,PL)))))
             = 1 - (1 - 0.7 \times (1 - (1 - 0.7)(1 - 0.4)))(1 - 0.4 \times (1 - (1 - 0.3)(1 - 0)))
             = 1 - (1 - 0.7 \times (1 - 0.3 \times 0.6))(1 - 0.4 \times (1 - 0.7))
             = 1 - 0.426 \times 0.88
2.
          = 0.625
```

Solution:

3. 0.982

```
0.70:: alice .
0.40::bob .
0.90:: charlie .

0.70:: alice_AI .
0.40:: alice_PL .
0.30::bob_AI .

out1 :- \+(\+alice_\+bob_\+charlie) .
```

query (out1).

```
Solution:
```

4. 0.5512

```
0.70:: alice .
0.40:: bob .
0.90:: charlie .

0.70:: alice_AI .
0.40:: alice_PL .
0.30:: bob_AI .

p1 :- bob, bob_AI .

p2 :- alice , alice_AI .
out2 :- \+(\+p1,\+p2) .

query(out2) .
```

Solution:

5. (3.) 0.62512

```
0.70:: alice .
0.40::bob .
0.90:: charlie .

0.70:: alice_AI .
0.40:: alice_PL .
0.30::bob_AI .

p1 :- alice , alice_AI .
p2 :- alice , alice_PL .
p3 :- bob , bob_AI .

out3 :- \+(\+p1,\+p2,\+p3) .
query(out3) .
```