

# Inference for SRL Report

Capita Selecta AI (Probabilistic Programming) 2016-2017

Alexander Tang s0189509  
Daniel Andrés Pérez Pérez r0605947

Alex: proposal: change all  $e1, e2, e3$  to  $eh, em, en$  - D: I'd say let it like that to be consistence with the number indices of the other variables. I just add a small description about what is the meaning of each variable to be more precise

Alex: table IV: is this needed? The  $A \models B$  implicitly means  $A \models B$  and  $B \models A$ . D: I made this according to the ENC1 description of the paper, Since table `cnfRepresentationEnc1` is in essence the same as `enc1Encoding`, I took off `cnfRepresentationEnc1` from the table list

Alex: table V: maybe add  $W(\theta_{a2} - b2, e3) = 1$  just to cover everything? - I guess it's not need it, otherwise we need also to add  $\lambda_{b1} = 1, \lambda_{b1} = 2, \lambda_{a1} = 1, \lambda_{a2} = 1, etc.$

Alex: table VII: what's the difference with table VI? D: Actually they are the same, I wanted to explicitly note the CNF representation and the actual encoding but I ended up with the same tables. I took off table `cnfRepresentationEnc1`

Alex: table VIII: I'm not sure if  $W(ro\ e2) = 0.19/(1-0.01)$  ; is it not just 0.19? Because it's a parent node and does not have any conditionals. I may be very wrong on this; I'm not sure about this. - D: You are right, Earthquake is marginal independent of the others (there is no conditional dependency such that  $P(E \mid \text{SomeNodes})$ ), I just wanted to be explicit according with the method described in the paper, we can always get rid of the "unseless" operation

Daniel: Not sure if the 6th clause of the network variable A in IV is correctly substituted due to the property describe in the paper "Consider again Figure 1 and imagine that the parameter  $\theta_{c1|a1}$  were 0. Given that this parameter is known to be 0, all models that set this parameter variable to true will have weight 0."

Daniel: For VI, should we also chance the closure that contains a probability of 0, like we did in the ENC1 encoding? Look comment above

## I. PROBABILISTIC INFERENCE USING WEIGHTED MODEL COUNTING

### A. PGM to CNF

Table I shows the semantics of the domain variables used for those tasks.

Tables II and III show the logical variables used for encoded the Bayesian Network in [1].

Table I. RANDOM VARIABLES AND DOMAINS SEMANTICS

Variable	Domain
B = Burglary	b1 = theres is burglary
	b2 = theres is no burglary
E = Earthquake	a1 = there is heavyearthquake
	a2 = there is mild earthquake
	a3 = there is no earthquake
A = Alarm	a1 = alarm rings
	a2 = alarm does not ring
J = John	j1 = John calls
	j2 = John does not call
M = Mary	m1 = Mary calls
	m2 = Mary does not call

Table II. LOGICAL VARIABLES USING ENC1

Network variables	Indicator Variable	CTP
B	$\lambda_{b1}, \lambda_{b2}$	$\theta_{b1}, \theta_{b2}$
E	$\lambda_{e1}, \lambda_{e2}, \lambda_3$	$\theta_{e1}, \theta_{e2}, \theta_3$
A	$\lambda_{a1}, \lambda_{a2}$	$\theta_{a1 b1,e1}, \theta_{a1 b1,e2}, \theta_{a1 b1,e3},$ $\theta_{a1 b2,e1}, \theta_{a1 b2,e2}, \theta_{a1 b2,e3},$ $\theta_{a2 b1,e1}, \theta_{a2 b1,e2}, \theta_{a2 b1,e3},$ $\theta_{a2 b2,e1}, \theta_{a2 b2,e2}, \theta_{a2 b2,e3}$
J	$\lambda_{j1}, \lambda_{j2}$	$\theta_{j1 a1}, \theta_{j2 a1}, \theta_{j1 a2}, \theta_{j2 a2}$
M	$\lambda_{m1}, \lambda_{m2}$	$\theta_{m1 a1}, \theta_{m2 a1}, \theta_{m1 a2}, \theta_{m2 a2}$

Table III. LOGICAL VARIABLES USING ENC2

Variables	Indicator Variable	CTP
B	$\lambda_{b1}, \lambda_{b2}$	$\rho_{b1}$
E	$\lambda_{e1}, \lambda_{e2}, \lambda_3$	$\rho_{e1}, \rho_{e2}$
A	$\lambda_{a1}, \lambda_{a2}$	$\rho_{a1 b1,c1}, \rho_{a1 b1,c2}, \rho_{a1 b1,c3},$ $\rho_{a1 b2,c1}, \rho_{a1 b2,c2}, \rho_{a1 b2,c3}$
J	$\lambda_{j1}, \lambda_{j2}$	$\rho_{j1 a1}, \rho_{j1 a2}$
M	$\lambda_{m1}, \lambda_{m2}$	$\rho_{m1 a1}, \rho_{m1 a2}$

Table IV represents the encoded Bayesian Network using ENC1 and table V contains the corresponding weights.

Likewise, table VI represent the encoded Bayesian Network using ENC2 and table VII contains the corresponding weights.

## REFERENCES

- [1] Bayesian networks.

Table IV. BAYESIAN NETWORK ENCODED USING ENC1

Network variable	Indicator Clauses	ENC1 Parameter Clauses
B	$\lambda_{b1} \vee \lambda_{b2},$ $\neg \lambda_{b1} \vee \neg \lambda_{b2}$	$\lambda_{b1} \Rightarrow \theta_{b1}, \theta_{b1} \Rightarrow \lambda_{b1}$ $\lambda_{b2} \Rightarrow \theta_{b2}, \theta_{b2} \Rightarrow \lambda_{b2}$
E	$\lambda_{e1} \vee \lambda_{e2} \vee \lambda_{e3},$ $\neg \lambda_{e1} \vee \neg \lambda_{e2},$ $\neg \lambda_{e1} \vee \neg \lambda_{e3},$ $\neg \lambda_{e2} \vee \neg \lambda_{e3}$	$\lambda_{e1} \Rightarrow \theta_{e1}, \theta_{e1} \Rightarrow \lambda_{e1}$ $\lambda_{e2} \Rightarrow \theta_{e2}, \theta_{e2} \Rightarrow \lambda_{e2}$ $\lambda_{e3} \Rightarrow \theta_{e3}, \theta_{e3} \Rightarrow \lambda_{e3}$
A	$\lambda_{a1} \vee \lambda_{a2},$ $\neg \lambda_{a1} \vee \neg \lambda_{a2}$	$\lambda_{a1} \wedge \lambda_{b1} \wedge \lambda_{e1} \Rightarrow \theta_{a1 b1,e1}, \theta_{a1 b1,e1} \Rightarrow \lambda_{a1}, \theta_{a1 b1,e1} \Rightarrow \lambda_{b1}, \theta_{a1 b1,e1} \Rightarrow \lambda_{e1}$ $\lambda_{a1} \wedge \lambda_{b1} \wedge \lambda_{e2} \Rightarrow \theta_{a1 b1,e2}, \theta_{a1 b1,e2} \Rightarrow \lambda_{a1}, \theta_{a1 b1,e2} \Rightarrow \lambda_{b1}, \theta_{a1 b1,e2} \Rightarrow \lambda_{e2}$ $\lambda_{a1} \wedge \lambda_{b1} \wedge \lambda_{e3} \Rightarrow \theta_{a1 b1,e3}, \theta_{a1 b1,e3} \Rightarrow \lambda_{a1}, \theta_{a1 b1,e3} \Rightarrow \lambda_{b1}, \theta_{a1 b1,e3} \Rightarrow \lambda_{e3}$ $\lambda_{a1} \wedge \lambda_{b2} \wedge \lambda_{e1} \Rightarrow \theta_{a1 b2,e1}, \theta_{a1 b2,e1} \Rightarrow \lambda_{a1}, \theta_{a1 b2,e1} \Rightarrow \lambda_{b2}, \theta_{a1 b2,e1} \Rightarrow \lambda_{e1}$ $\lambda_{a1} \wedge \lambda_{b2} \wedge \lambda_{e2} \Rightarrow \theta_{a1 b2,e2}, \theta_{a1 b2,e2} \Rightarrow \lambda_{a1}, \theta_{a1 b2,e2} \Rightarrow \lambda_{b2}, \theta_{a1 b2,e2} \Rightarrow \lambda_{e2}$ $\neg \lambda_{a1} \vee \neg \lambda_{b2} \vee \neg \lambda_{e3}$ $\lambda_{a2} \wedge \lambda_{b1} \wedge \lambda_{e1} \Rightarrow \theta_{a2 b1,e1}, \theta_{a2 b1,e1} \Rightarrow \lambda_{a2}, \theta_{a2 b1,e1} \Rightarrow \lambda_{b1}, \theta_{a2 b1,e1} \Rightarrow \lambda_{e1}$ $\lambda_{a2} \wedge \lambda_{b1} \wedge \lambda_{e2} \Rightarrow \theta_{a2 b1,e2}, \theta_{a2 b1,e2} \Rightarrow \lambda_{a2}, \theta_{a2 b1,e2} \Rightarrow \lambda_{b1}, \theta_{a2 b1,e2} \Rightarrow \lambda_{e2}$ $\lambda_{a2} \wedge \lambda_{b1} \wedge \lambda_{e3} \Rightarrow \theta_{a2 b1,e3}, \theta_{a2 b1,e3} \Rightarrow \lambda_{a2}, \theta_{a2 b1,e3} \Rightarrow \lambda_{b1}, \theta_{a2 b1,e3} \Rightarrow \lambda_{e3}$ $\lambda_{a2} \wedge \lambda_{b2} \wedge \lambda_{e1} \Rightarrow \theta_{a2 b2,e1}, \theta_{a2 b2,e1} \Rightarrow \lambda_{a2}, \theta_{a2 b2,e1} \Rightarrow \lambda_{b2}, \theta_{a2 b2,e1} \Rightarrow \lambda_{e1}$ $\lambda_{a2} \wedge \lambda_{b2} \wedge \lambda_{e2} \Rightarrow \theta_{a2 b2,e2}, \theta_{a2 b2,e2} \Rightarrow \lambda_{a2}, \theta_{a2 b2,e2} \Rightarrow \lambda_{b2}, \theta_{a2 b2,e2} \Rightarrow \lambda_{e2}$ $\lambda_{a2} \wedge \lambda_{b2} \wedge \lambda_{e3} \Rightarrow \theta_{a2 b2,e3}, \theta_{a2 b2,e3} \Rightarrow \lambda_{a2}, \theta_{a2 b2,e3} \Rightarrow \lambda_{b2}, \theta_{a2 b2,e3} \Rightarrow \lambda_{e3}$
J	$\lambda_{j1} \vee \lambda_{j2},$ $\neg \lambda_{j1} \vee \neg \lambda_{j2}$	$\lambda_{j1} \wedge \lambda_{a1} \Rightarrow \theta_{j1 a1}, \theta_{j1 a1} \Rightarrow \lambda_{j1}, \theta_{j1 a1} \Rightarrow \lambda_{a1}$ $\lambda_{j1} \wedge \lambda_{a2} \Rightarrow \theta_{j1 a2}, \theta_{j1 a2} \Rightarrow \lambda_{j1}, \theta_{j1 a2} \Rightarrow \lambda_{a2}$ $\lambda_{j2} \wedge \lambda_{a1} \Rightarrow \theta_{j2 a1}, \theta_{j2 a1} \Rightarrow \lambda_{j2}, \theta_{j2 a1} \Rightarrow \lambda_{a1}$ $\lambda_{j2} \wedge \lambda_{a2} \Rightarrow \theta_{j2 a2}, \theta_{j2 a2} \Rightarrow \lambda_{j2}, \theta_{j2 a2} \Rightarrow \lambda_{a2}$
M	$\lambda_{m1} \vee \lambda_{m2},$ $\neg \lambda_{m1} \vee \neg \lambda_{m2}$	$\lambda_{m1} \wedge \lambda_{a1} \Rightarrow \theta_{m1 a1}, \theta_{m1 a1} \Rightarrow \lambda_{m1}, \theta_{m1 a1} \Rightarrow \lambda_{a1}$ $\lambda_{m1} \wedge \lambda_{a2} \Rightarrow \theta_{m1 a2}, \theta_{m1 a2} \Rightarrow \lambda_{m1}, \theta_{m1 a2} \Rightarrow \lambda_{a2}$ $\lambda_{m2} \wedge \lambda_{a1} \Rightarrow \theta_{m2 a1}, \theta_{m2 a1} \Rightarrow \lambda_{m2}, \theta_{m2 a1} \Rightarrow \lambda_{a1}$ $\lambda_{m2} \wedge \lambda_{a2} \Rightarrow \theta_{m2 a2}, \theta_{m2 a2} \Rightarrow \lambda_{m2}, \theta_{m2 a2} \Rightarrow \lambda_{a2}$

Table V. WEIGHTS ASSOCIATION USING ENC1 WHERE MISSING WEIGHTS ARE SET TO ONE

Weights	Value
$W(\theta_{b1})$	0.7
$W(\theta_{b2})$	0.3
$W(\theta_{e1})$	0.01
$W(\theta_{e2})$	0.19
$W(\theta_{e3})$	0.80
$W(\theta_{a1 b1,e1})$	0.90
$W(\theta_{a1 b1,e2})$	0.85
$W(\theta_{a1 b1,e3})$	0.80
$W(\theta_{a1 b2,e1})$	0.30
$W(\theta_{a1 b2,e2})$	0.10
$W(\theta_{a1 b2,e3})$	0
$W(\theta_{a2 b1,e1})$	0.10
$W(\theta_{a2 b1,e2})$	0.15
$W(\theta_{a2 b1,e3})$	0.20
$W(\theta_{a2 b2,e1})$	0.70
$W(\theta_{a2 b2,e2})$	0.90
$W(\theta_{j1 a1})$	0.80
$W(\theta_{j1 a2})$	0.10
$W(\theta_{j2 a1})$	0.20
$W(\theta_{j2 a2})$	0.90
$W(\theta_{m1 a1})$	0.80
$W(\theta_{m1 a2})$	0.10
$W(\theta_{m2 a1})$	0.20
$W(\theta_{m2 a2})$	0.90

Table VI. BAYESIAN NETWORK ENCODED USING ENC2

Network variable	Indicator Clauses	ENC1 Parameter Clauses
B	$\lambda_{b1} \vee \lambda_{b2},$ $\neg\lambda_{b1} \vee \neg\lambda_{b2}$	$\rho_{b1} \Rightarrow \lambda_{b1}$ $\neg\rho_{b1} \Rightarrow \lambda_{b2}$
E	$\lambda_{e1} \vee \lambda_{e2} \vee \lambda_{e3},$ $\neg\lambda_{e1} \vee \neg\lambda_{e2},$ $\neg\lambda_{e1} \vee \neg\lambda_{e3},$ $\neg\lambda_{e2} \vee \neg\lambda_{e3}$	$\rho_{e1} \Rightarrow \lambda_{e1}$ $\neg\rho_{e1} \wedge \rho_{e2} \Rightarrow \lambda_{e2}$ $\neg\rho_{e1} \wedge \neg\rho_{e2} \Rightarrow \lambda_{e3}$
A	$\lambda_{a1} \vee \lambda_{a2},$ $\neg\lambda_{a1} \vee \neg\lambda_{a2}$	$\lambda_{b1} \wedge \lambda_{e1} \wedge \rho_{a1 b1,e1} \Rightarrow \lambda_{a1}$ $\lambda_{b1} \wedge \lambda_{e2} \wedge \rho_{a1 b1,e2} \Rightarrow \lambda_{a1}$ $\lambda_{b1} \wedge \lambda_{e3} \wedge \rho_{a1 b1,e3} \Rightarrow \lambda_{a1}$ $\lambda_{b2} \wedge \lambda_{e1} \wedge \rho_{a1 b2,e1} \Rightarrow \lambda_{a1}$ $\lambda_{b2} \wedge \lambda_{e2} \wedge \rho_{a1 b2,e2} \Rightarrow \lambda_{a1}$ $\lambda_{b2} \wedge \lambda_{e3} \wedge \rho_{a1 b2,e3} \Rightarrow \lambda_{a1}$ $\lambda_{b1} \wedge \lambda_{e1} \wedge \neg\rho_{a1 b1,e1} \Rightarrow \lambda_{a2}$ $\lambda_{b1} \wedge \lambda_{e2} \wedge \neg\rho_{a1 b1,e2} \Rightarrow \lambda_{a2}$ $\lambda_{b1} \wedge \lambda_{e3} \wedge \neg\rho_{a1 b1,e3} \Rightarrow \lambda_{a2}$ $\lambda_{b2} \wedge \lambda_{e1} \wedge \neg\rho_{a1 b2,e1} \Rightarrow \lambda_{a2}$ $\lambda_{b2} \wedge \lambda_{e2} \wedge \neg\rho_{a1 b2,e2} \Rightarrow \lambda_{a2}$ $\lambda_{b2} \wedge \lambda_{e3} \wedge \neg\rho_{a1 b2,e3} \Rightarrow \lambda_{a2}$
J	$\lambda_{j1} \vee \lambda_{j2},$ $\neg\lambda_{j1} \vee \neg\lambda_{j2}$	$\lambda_{a1} \wedge \rho_{j1 a1} \Rightarrow \lambda_{j1}$ $\lambda_{a2} \wedge \rho_{j1 a2} \Rightarrow \lambda_{j1}$ $\lambda_{a1} \wedge \neg\rho_{j1 a1} \Rightarrow \lambda_{j2}$ $\lambda_{a2} \wedge \neg\rho_{j1 a2} \Rightarrow \lambda_{j2}$
M	$\lambda_{m1} \vee \lambda_{m2},$ $\neg\lambda_{m1} \vee \neg\lambda_{m2}$	$\lambda_{a1} \wedge \rho_{m1 a1} \Rightarrow \lambda_{m1}$ $\lambda_{a2} \wedge \rho_{m1 a2} \Rightarrow \lambda_{m1}$ $\lambda_{a1} \wedge \neg\rho_{m1 a1} \Rightarrow \lambda_{m2}$ $\lambda_{a2} \wedge \neg\rho_{m1 a2} \Rightarrow \lambda_{m2}$

Table VII. WEIGHTS ASSOCIATION USING ENC2 WHERE MISSING WEIGHTS ARE SET TO ONE

Weights	Value
$W(\rho_{b1})$	0.7
$W(\neg\rho_{b1})$	0.3
$W(\rho_{e1})$	0.01
$W(\rho_{e2})$	$0.19/(1-0.01) = 0.19$
$W(\neg\rho_{e1})$	$1-0.01 = 0.99$
$W(\neg\rho_{e2})$	$1-0.19 = 0.81$
$W(\rho_{a1 b1,e1})$	0.90
$W(\neg\rho_{a1 b1,e1})$	$1-0.90=0.10$
$W(\rho_{a1 b1,e2})$	0.85
$W(\neg\rho_{a1 b1,e2})$	$1-0.85=0.15$
$W(\rho_{a1 b1,e3})$	0.80
$W(\neg\rho_{a1 b1,e3})$	$1-0.80=0.20$
$W(\rho_{a1 b2,e1})$	0.30
$W(\neg\rho_{a1 b2,e1})$	$1-0.30=0.70$
$W(\rho_{a1 b2,e2})$	0.10
$W(\neg\rho_{a1 b2,e2})$	$1-0-10=0.90$
$W(\rho_{a1 b2,e3})$	0
$W(\neg\rho_{a1 b2,e3})$	$1-0=1$
$W(\rho_{j1 a1})$	0.80
$W(\neg\rho_{j1 a1})$	$1-0.80=0.20$
$W(\rho_{j1 a2})$	0.10
$W(\neg\rho_{j1 a2})$	$1-0.10=0.90$