

# Avoiding Repetition in Repeated Inference on Probabilistic Relational Models: The Lifted Junction Tree Algorithm – Evaluation Setting

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## 1. Input Description for the Empirical Evaluation

During the empirical evaluation, we vary the following, (i) largest domain size  $n$ , (ii) number of parclusters  $n_J$ , and (iii) lifted width  $(w_g, w_\#)$  in which  $w_g$  is the ground width and  $w_\#$  the counting width. In the following, we provide a description of the basic model as well as the variations w.r.t.  $n$ ,  $n_J$ ,  $w_g$ , and  $w_\#$ , the evidence setup, and the fusion test model.

## 2. Basic Setting

The basic input model is  $G_{ex}$  with boolean ranges,  $n = 1000$ , which we use for each logvar,  $n_J = 3$ ,  $w_g = 3$ , and  $w_\# = 1$ . Explicitly, we have  $\mathbf{L} = \{D, W, M, X\}$ ,  $\Phi = \{\phi_0, \phi_1, \phi_2, \phi_3\}$ , and  $\mathbf{R} = \{Epid, Nat, Man, Sick, Travel, Treat\}$ . The domain sizes are  $n = |\mathcal{D}(D)| = |\mathcal{D}(W)| = |\mathcal{D}(M)| = |\mathcal{D}(X)| = 1000$ . The PRVs are  $Epid$ ,  $Sick(X)$ ,  $Nat(D)$ ,  $Man(W)$ ,  $Travel(X)$ , and  $Treat(X, M)$  with boolean range. The parfactors in  $G_{ex}$  are  $g_0 = \phi_0(Epid)$ ,  $g_1 = \phi_1(Epid, Nat(D), Man(W))$ ,  $g_2 = \phi_2(Epid, Sick(X), Travel(X))$ , and  $g_3 = \phi_3(Epid, Sick(X), Treat(X, M))$  with random potentials. When FOKC is involved, evidence is empty as to not put FOKC at a disadvantage. Otherwise, there exists observations for 10% of the instances of each one-logvar PRV.

## 3. Varying the Domain Size

$G_{ex}$  remains the same, while  $n = |\mathcal{D}(D)| = |\mathcal{D}(W)| = |\mathcal{D}(M)| = |\mathcal{D}(X)|$  varies. The last line holds the basic setting.  $\max |\mathcal{A}|$  refers to the largest number of arguments.

Table 1: Key numbers about the input models used when varying  $n$  without FOKC

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ \mathbf{G} $	$ gr(\mathbf{G}) $	$ G_i $ in	$\max  \mathcal{A} $
5	3	3	1	4	4	11	{1, 2}	3
10	3	3	1	8	4	211	{1, 2}	3
15	3	3	1	12	4	301	{1, 2}	3
20	3	3	1	16	4	821	{1, 2}	3
50	3	3	1	40	4	5051	{1, 2}	3
100	3	3	1	80	4	20101	{1, 2}	3

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
150	3	3	1	120	4	45151	{1, 2}	3
200	3	3	1	160	4	80201	{1, 2}	3
250	3	3	1	200	4	125251	{1, 2}	3
300	3	3	1	240	4	180301	{1, 2}	3
350	3	3	1	280	4	245351	{1, 2}	3
400	3	3	1	320	4	320401	{1, 2}	3
450	3	3	1	360	4	405451	{1, 2}	3
500	3	3	1	400	4	500501	{1, 2}	3
550	3	3	1	440	4	605551	{1, 2}	3
600	3	3	1	480	4	720601	{1, 2}	3
650	3	3	1	520	4	845651	{1, 2}	3
700	3	3	1	560	4	980701	{1, 2}	3
750	3	3	1	600	4	1125751	{1, 2}	3
800	3	3	1	640	4	1280801	{1, 2}	3
850	3	3	1	680	4	1445851	{1, 2}	3
900	3	3	1	720	4	1620901	{1, 2}	3
950	3	3	1	760	4	1805951	{1, 2}	3
1000	3	3	1	800	4	2001001	{1, 2}	3

Table 3: Key numbers about the input models used when varying  $n$  with FOKC

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
2	3	3	1	0	4	11	{1, 2}	3
4	3	3	1	0	4	37	{1, 2}	3
6	3	3	1	0	4	79	{1, 2}	3
8	3	3	1	0	4	137	{1, 2}	3
10	3	3	1	0	4	211	{1, 2}	3
12	3	3	1	0	4	301	{1, 2}	3
14	3	3	1	0	4	407	{1, 2}	3
16	3	3	1	0	4	529	{1, 2}	3
18	3	3	1	0	4	667	{1, 2}	3
20	3	3	1	0	4	821	{1, 2}	3
50	3	3	1	0	4	5051	{1, 2}	3
100	3	3	1	0	4	20101	{1, 2}	3
150	3	3	1	0	4	45151	{1, 2}	3
200	3	3	1	0	4	80201	{1, 2}	3
250	3	3	1	0	4	125251	{1, 2}	3
300	3	3	1	0	4	180301	{1, 2}	3
350	3	3	1	0	4	245351	{1, 2}	3
400	3	3	1	0	4	320401	{1, 2}	3
450	3	3	1	0	4	405451	{1, 2}	3
500	3	3	1	0	4	500501	{1, 2}	3

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
550	3	3	1	0	4	605551	{1,2}	3
600	3	3	1	0	4	720601	{1,2}	3
650	3	3	1	0	4	845651	{1,2}	3
700	3	3	1	0	4	980701	{1,2}	3
750	3	3	1	0	4	1125751	{1,2}	3
800	3	3	1	0	4	1280801	{1,2}	3
850	3	3	1	0	4	1445851	{1,2}	3
900	3	3	1	0	4	1620901	{1,2}	3
950	3	3	1	0	4	1805951	{1,2}	3
1000	3	3	1	0	4	2001001	{1,2}	3
2000	3	3	1	0	4	2001001	{1,2}	3
3000	3	3	1	0	4	2001001	{1,2}	3
4000	3	3	1	0	4	2001001	{1,2}	3
5000	3	3	1	0	4	2001001	{1,2}	3
6000	3	3	1	0	4	2001001	{1,2}	3
7000	3	3	1	0	4	2001001	{1,2}	3
8000	3	3	1	0	4	2001001	{1,2}	3
9000	3	3	1	0	4	2001001	{1,2}	3
10000	3	3	1	0	4	2001001	{1,2}	3

In the following, the tables show two values in the  $\mathbf{E}$  column for number of observations, 0 for runs with FOKC and the other values for runs without FOKC.

#### 4. Varying the Number of Parclusters

When varying  $n_J$  from 2 to 11,  $w_g$  and  $w_\#$  remain fixed. Each new parcluster has a lifted width of (3, 0), leaving the overall lifted width at (3, 1).  $G_{ex}$  appears where  $n_J = 3$  and  $n = 1000$ . A small jump in  $|gr(G)|$  occurs when additional PRVs trigger a new parcluster. A larger jump occurs if an additional logvar triggers a new parcluster.

Table 5: Key numbers about the input models used when varying  $n_J$

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
10	2	3	1	8,0	3	111	{1,2}	3
10	3	3	1	8,0	4	211	{1,2}	3
10	4	3	1	10,0	5	221	{1,2}	3
10	5	3	1	12,0	6	231	{1,2}	3
10	6	3	1	12,0	7	331	{1,2}	3
10	7	3	1	14,0	8	341	{1,2}	3
10	8	3	1	14,0	9	441	{1,2}	3
10	9	3	1	16,0	10	451	{1,2}	3
10	10	3	1	18,0	11	461	{1,2}	3
10	11	3	1	18,0	12	561	{1,2}	3

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
100	2	3	1	80,0	3	10101	{1,2}	3
100	3	3	1	80,0	4	20101	{1,2}	3
100	4	3	1	100,0	5	20201	{1,2}	3
100	5	3	1	120,0	6	20301	{1,2}	3
100	6	3	1	120,0	7	30301	{1,2}	3
100	7	3	1	140,0	8	30401	{1,2}	3
100	8	3	1	140,0	9	40401	{1,2}	3
100	9	3	1	160,0	10	40501	{1,2}	3
100	10	3	1	180,0	11	40601	{1,2}	3
100	11	3	1	180,0	12	50601	{1,2}	3
1000	2	3	1	800,0	3	1001001	{1,2}	3
1000	3	3	1	800,0	4	2001001	{1,2}	3
1000	4	3	1	1000,0	5	2002001	{1,2}	3
1000	5	3	1	1200,0	6	2003001	{1,2}	3
1000	6	3	1	1200,0	7	3003001	{1,2}	3
1000	7	3	1	1400,0	8	3004001	{1,2}	3
1000	8	3	1	1400,0	9	4004001	{1,2}	3
1000	9	3	1	1600,0	10	4005001	{1,2}	3
1000	10	3	1	1800,0	11	4006001	{1,2}	3
1000	11	3	1	1800,0	12	5006001	{1,2}	3

## 5. Varying the Ground Width

Varying  $w_g$  from 2 to 11 means that  $n_J$  and  $w_\#$  appear fixed and each parcluster has a ground width of  $w_g$ .  $G_{ex}$  appears where  $w_g = 3$  and  $n = 1000$ .

Table 7: Key numbers about the input models used when varying  $w_g$

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
10	3	2	1	6,0	4	31	{2}	2
10	3	3	1	8,0	4	121	{1,2}	3
10	3	4	1	12,0	10	631	{3,4}	3
10	3	5	1	12,0	16	1231	{5,7}	3
10	3	6	1	18,0	25	2131	{8,9}	3
10	3	7	1	24,0	34	2761	{11,12}	3
10	3	8	1	24,0	43	3661	{14,15}	3
10	3	9	1	24,0	52	4561	{17,18}	3
10	3	10	1	30,0	61	5191	{20,21}	3
10	3	11	1	36,0	70	5821	{23,24}	3
100	3	2	1	60,0	4	301	{2}	2
100	3	3	1	80,0	4	10201	{1,2}	3
100	3	4	1	120,0	10	60301	{3,4}	3
100	3	5	1	120,0	16	120301	{5,7}	3

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
100	3	6	1	180,0	25	210301	{8, 9}	3
100	3	7	1	240,0	34	270601	{11, 12}	3
100	3	8	1	240,0	43	360601	{14, 15}	3
100	3	9	1	240,0	52	450601	{17, 18}	3
100	3	10	1	300,0	61	510901	{20, 21}	3
100	3	11	1	360,0	70	571201	{23, 24}	3
1000	3	2	1	600,0	4	3001	{2}	2
1000	3	3	1	800,0	4	1002001	{1, 2}	3
1000	3	4	1	1200,0	10	6003001	{3, 4}	3
1000	3	5	1	1200,0	16	12003001	{5, 7}	3
1000	3	6	1	1800,0	25	21003001	{8, 9}	3
1000	3	7	1	2400,0	34	27006001	{11, 12}	3
1000	3	8	1	2400,0	43	36006001	{14, 15}	3
1000	3	9	1	2400,0	52	45006001	{17, 18}	3
1000	3	10	1	3000,0	61	51009001	{20, 21}	3
1000	3	11	1	3600,0	70	57012001	{23, 24}	3

## 6. Varying the Counting Width

When varying  $w_\#$  from 0 to 9,  $n_J$  and  $w_g$  are fixed and each of the three parclusters has a counting width of  $w_\#$ .  $G_{ex}$  appears where  $w_\# = 1$  and  $n = 1000$ .

Table 9: Key numbers about the input models used when varying  $w_\#$

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
10	3	3	0	6,0	4	31	{2}	2
10	3	3	1	12,0	4	301	{1, 2}	3
10	3	3	2	18,0	4	3001	{1, 2}	4
10	3	3	3	24,0	7	6001	{2, 3}	4
10	3	3	4	30,0	10	9001	{3, 4}	4
10	3	3	5	36,0	13	12001	{4, 5}	4
10	3	3	6	42,0	16	15001	{5, 6}	4
10	3	3	7	48,0	19	18001	{6, 7}	4
10	3	3	8	54,0	22	21001	{7, 8}	4
10	3	3	9	60,0	25	24001	{8, 9}	4
100	3	3	0	60,0	4	301	{2}	2
100	3	3	1	120,0	4	30001	{1, 2}	3
100	3	3	2	180,0	4	3000001	{1, 2}	4
100	3	3	3	240,0	7	6000001	{2, 3}	4
100	3	3	4	300,0	10	9000001	{3, 4}	4
100	3	3	5	360,0	13	12000001	{4, 5}	4
100	3	3	6	420,0	16	15000001	{5, 6}	4
100	3	3	7	480,0	19	18000001	{6, 7}	4

$n$	$n_J$	$w_g$	$w_{\#}$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
100	3	3	8	540,0	22	21000001	{7, 8}	4
100	3	3	9	600,0	25	24000001	{8, 9}	4
1000	3	3	0	600,0	4	3001	{2}	2
1000	3	3	1	1200,0	4	3,000,001	{1, 2}	3
1000	3	3	2	1800,0	10	$3 \cdot 1000^3 + 1$	{1, 2}	4
1000	3	3	3	2400,0	16	$6 \cdot 1000^3 + 1$	{2, 3}	4
1000	3	3	4	3000,0	25	$9 \cdot 1000^3 + 1$	{3, 4}	4
1000	3	3	5	3600,0	34	$12 \cdot 1000^3 + 1$	{4, 5}	4
1000	3	3	6	4200,0	43	$15 \cdot 1000^3 + 1$	{5, 6}	4
1000	3	3	7	4800,0	52	$18 \cdot 1000^3 + 1$	{6, 7}	4
1000	3	3	8	5400,0	61	$21 \cdot 1000^3 + 1$	{7, 8}	4
1000	3	3	9	6000,0	70	$24 \cdot 1000^3 + 1$	{8, 9}	4

## 7. Varying the Evidence Coverage

For evidence testing, we add evidence for the 1-logvar PRVs in  $G_{ex}$ , i.e., four evidence parafactors, in 10% steps. The line with  $n = 1000$  and  $|\mathbf{E}| = 0$  is  $G_{ex}$ .

Table 11: Key numbers about the input models used when varying evidence coverage

$n$	$n_J$	$w_g$	$w_{\#}$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
10	3	3	1	0	4	211	{1, 2}	3
10	3	3	1	4	4	211	{1, 2}	3
10	3	3	1	8	4	211	{1, 2}	3
10	3	3	1	12	4	211	{1, 2}	3
10	3	3	1	16	4	211	{1, 2}	3
10	3	3	1	20	4	211	{1, 2}	3
10	3	3	1	24	4	211	{1, 2}	3
10	3	3	1	28	4	211	{1, 2}	3
10	3	3	1	32	4	211	{1, 2}	3
10	3	3	1	36	4	211	{1, 2}	3
10	3	3	1	36	4	211	{1, 2}	3
100	3	3	1	0	4	20101	{1, 2}	3
100	3	3	1	40	4	20101	{1, 2}	3
100	3	3	1	80	4	20101	{1, 2}	3
100	3	3	1	120	4	20101	{1, 2}	3
100	3	3	1	160	4	20101	{1, 2}	3
100	3	3	1	200	4	20101	{1, 2}	3
100	3	3	1	240	4	20101	{1, 2}	3
100	3	3	1	280	4	20101	{1, 2}	3
100	3	3	1	320	4	20101	{1, 2}	3
100	3	3	1	360	4	20101	{1, 2}	3
100	3	3	1	396	4	20101	{1, 2}	3

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $ in	$\max  \mathcal{A} $
1000	3	3	1	0	4	2001001	{1, 2}	3
1000	3	3	1	400	4	2001001	{1, 2}	3
1000	3	3	1	800	4	2001001	{1, 2}	3
1000	3	3	1	1200	4	2001001	{1, 2}	3
1000	3	3	1	1600	4	2001001	{1, 2}	3
1000	3	3	1	2000	4	2001001	{1, 2}	3
1000	3	3	1	2400	4	2001001	{1, 2}	3
1000	3	3	1	2800	4	2001001	{1, 2}	3
1000	3	3	1	3200	4	2001001	{1, 2}	3
1000	3	3	1	3600	4	2001001	{1, 2}	3
1000	3	3	1	3996	4	2001001	{1, 2}	3

## 8. Unnecessary Groundings and Fusion

For fusion, we use a slight variation of  $G_{ex}$  with  $n_J = 4$ , which has unnecessary groundings with LJT. After fusion,  $n_J = 3$  and  $w_g = 5$ . We vary the domain size  $n$  again.

Table 13: Key numbers about the input models used for evaluating fusion

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $	$\max  \mathcal{A} $
2	4 → 3	3 → 5	1	0	5	15	{1, 2} → {2, 3}	3
4	4 → 3	3 → 5	1	0	5	53	{1, 2} → {2, 3}	3
6	4 → 3	3 → 5	1	0	5	115	{1, 2} → {2, 3}	3
8	4 → 3	3 → 5	1	0	5	201	{1, 2} → {2, 3}	3
10	4 → 3	3 → 5	1	0	5	311	{1, 2} → {2, 3}	3
12	4 → 3	3 → 5	1	0	5	445	{1, 2} → {2, 3}	3
14	4 → 3	3 → 5	1	0	5	603	{1, 2} → {2, 3}	3
16	4 → 3	3 → 5	1	0	5	785	{1, 2} → {2, 3}	3
18	4 → 3	3 → 5	1	0	5	991	{1, 2} → {2, 3}	3
20	4 → 3	3 → 5	1	0	5	1221	{1, 2} → {2, 3}	3
50	4 → 3	3 → 5	1	0	5	7551	{1, 2} → {2, 3}	3
100	4 → 3	3 → 5	1	0	5	30101	{1, 2} → {2, 3}	3
150	4 → 3	3 → 5	1	0	5	67651	{1, 2} → {2, 3}	3
200	4 → 3	3 → 5	1	0	5	120201	{1, 2} → {2, 3}	3
250	4 → 3	3 → 5	1	0	5	187751	{1, 2} → {2, 3}	3
300	4 → 3	3 → 5	1	0	5	270301	{1, 2} → {2, 3}	3
350	4 → 3	3 → 5	1	0	5	367851	{1, 2} → {2, 3}	3
400	4 → 3	3 → 5	1	0	5	480401	{1, 2} → {2, 3}	3
450	4 → 3	3 → 5	1	0	5	607951	{1, 2} → {2, 3}	3
500	4 → 3	3 → 5	1	0	5	750501	{1, 2} → {2, 3}	3
550	4 → 3	3 → 5	1	0	5	908051	{1, 2} → {2, 3}	3
600	4 → 3	3 → 5	1	0	5	1080601	{1, 2} → {2, 3}	3
650	4 → 3	3 → 5	1	0	5	1268151	{1, 2} → {2, 3}	3

$n$	$n_J$	$w_g$	$w_\#$	$ \mathbf{E} $	$ G $	$ gr(G) $	$ G_i $	$\max  \mathcal{A} $
700	$4 \rightarrow 3$	$3 \rightarrow 5$	1	0	5	1470701	$\{1, 2\} \rightarrow \{2, 3\}$	3
750	$4 \rightarrow 3$	$3 \rightarrow 5$	1	0	5	1688251	$\{1, 2\} \rightarrow \{2, 3\}$	3
800	$4 \rightarrow 3$	$3 \rightarrow 5$	1	0	5	1920801	$\{1, 2\} \rightarrow \{2, 3\}$	3
850	$4 \rightarrow 3$	$3 \rightarrow 5$	1	0	5	2168351	$\{1, 2\} \rightarrow \{2, 3\}$	3
900	$4 \rightarrow 3$	$3 \rightarrow 5$	1	0	5	2430901	$\{1, 2\} \rightarrow \{2, 3\}$	3
950	$4 \rightarrow 3$	$3 \rightarrow 5$	1	0	5	2708451	$\{1, 2\} \rightarrow \{2, 3\}$	3
1000	$4 \rightarrow 3$	$3 \rightarrow 5$	1	0	5	3001001	$\{1, 2\} \rightarrow \{2, 3\}$	3

For more information regarding the input models, please send an email to:  
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