



# Lifted Inference Rules With Constraints

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## INTRODUCTION

### Markov Logic Networks (MLN) [Richardson & Domingos, 2006]

- A Markov Logic Network (MLN) is a set of pairs  $(F, w)$  where
  - $F$  is a formula in first order Logic
  - $w$  is a real number (weight of formula)
- When a world violates a formula, it becomes less probable, but not impossible.
- Together with a set of constants, it defines a Markov network.

$$\Delta_x = \Delta_y = \{1, 2, \dots, 10\}$$

$$1.5 \forall x \text{ Smokes}(x) \Rightarrow \text{Cancer}(x)$$

$$1.1 \forall x, y \text{ Smokes}(x) \wedge \text{Friends}(x, y) \Rightarrow \text{Smokes}(y)$$

#### Joint Probability :

$$P(y) = \frac{1}{Z} \exp \left( \sum_i w_i n_i(y) \right)$$

$$Z = \sum_{\text{assignment}} \exp \left( \sum_i w_i n_i(y) \right) \leftarrow \text{Intractable}$$

Weight of Formula  $i$       No. of true groundings of formula  $i$

- Lifted Inference** : Exploit symmetry in the network and infer about group of objects together. Ex : Lifted VE, Lifted BP, Lifting rules.
- In above example, probability of smoking of every person is same.
- Suppose we have evidence : Smokes(1), Friends(1,2).
- Evidence breaks symmetry, and imposes constraints on the values variables can take.
- How to do lifted inference when constraints are imposed on variables ?
- Focus of our work :
  - Propose a new constraint language which supports efficient lifted inference.
  - Modify lifted inference rules to work with our proposed constraint language.

## CONSTRAINT LANGUAGE (SetInEq)

- Weighted first order formulas with constraints.
- Three types of Atomic Constraints :

#### Subset Constraints

$$w \quad P(x) \Rightarrow Q(y, z), x \in \{A, B\}$$

#### Equality Constraints

$$w \quad P(x) \Rightarrow Q(y, z), x = y$$

#### Inequality Constraints

$$w \quad P(x) \Rightarrow Q(y, z), x \neq y$$

- Constraint Tuple : Conjunction of atomic constraints

$$w \quad P(x) \Rightarrow Q(y, z), x \in \{A, B\} \wedge y \notin \{C, D\} \wedge x \neq y$$

- Constraint Set : Disjunction of constraint tuples

$$w \quad P(x) \Rightarrow Q(y, z), (x \in \{A, B\} \wedge x \neq y) \vee (y \notin \{C, D\})$$

### Operations

- Join** : Conjunction of two constraint tuples .

- linear in the size of constraint tuples being joined.

- Project** : Eliminate a variable from a given constraint tuple.

- may require splitting when inequality constraints are present.
- Ex :  $x \neq y \wedge y \neq z$

### Canonical Representation

- Only one subset constraint per variable in a constraint tuple.

$$x \in \{A, B\} \wedge x \in \{B, C\} \wedge y \notin \{C, D\} \rightarrow x \in \{B\} \wedge y \notin \{C, D\}$$

- Identical supports for variables in eq/ineq constraints.

$$x \in \{A, B\} \wedge y \in \{A, B, C\} \wedge x = y \rightarrow x \in \{A, B\} \wedge y \in \{A, B\} \wedge x = y$$

#### Splitting Operation

$$x \in \{A, B, C\} \wedge y \in \{B, C, D\} \wedge x \neq y$$

$$x \in \{A\} \wedge y \in \{B, C\} \quad x \in \{A\} \wedge y \in \{D\} \quad x \in \{B, C\} \wedge y \in \{B, C\} \wedge x \neq y \quad x \in \{B, C\} \wedge y \in \{D\}$$

- Time Complexity to canonicalize a constraint set :  $O(mk + k^3)$ 
  - $m$  : no. of constants
  - $k$  : no. of variables

## LIFTING RULE 1 : DECOMPOSER RULE [Gogate & Domingos '11]

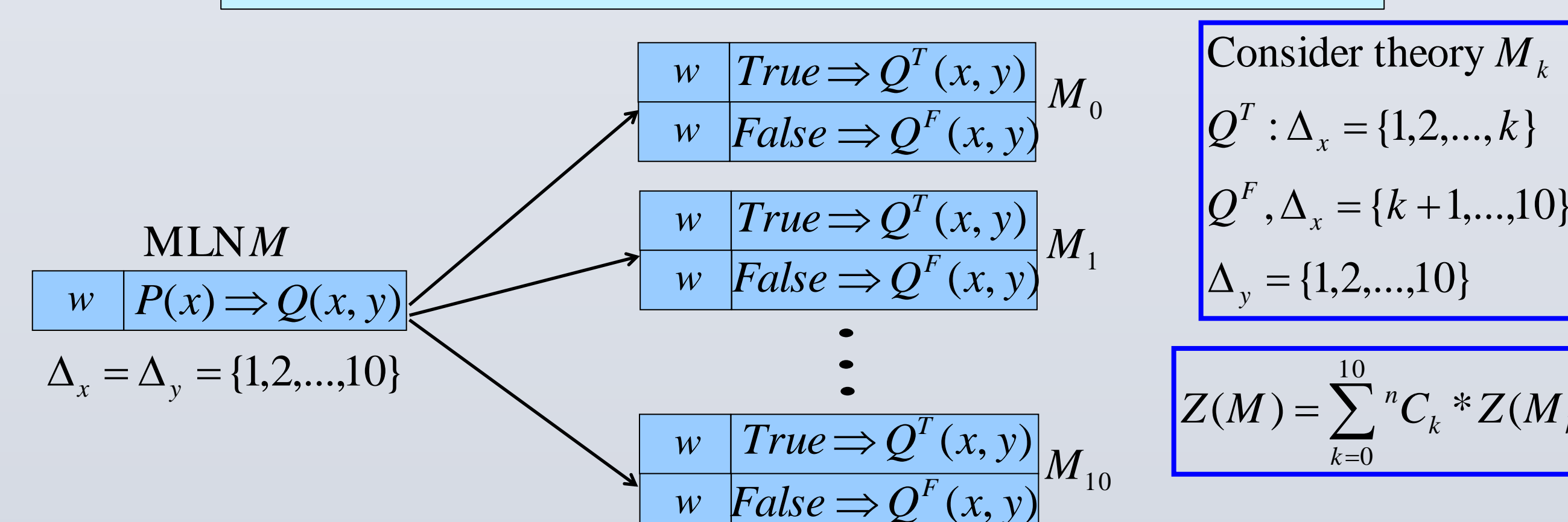
$$\begin{matrix} w & P(x) \Rightarrow Q(x, y) \\ x \in \{1, 2, 3\} \wedge y \in \{1, 2, 3\} \\ \vee \\ x \in \{1, 2\} \wedge y \in \{5, 6\} \end{matrix} \xrightarrow{\text{Decomposer Rule}} \begin{matrix} Z \left( \begin{matrix} w & P(x) \Rightarrow Q(x, y) \\ \Delta_x = \Delta_y = \{1, 2, \dots, 10\} \end{matrix} \right) \rightarrow \left( Z \left( \begin{matrix} w & P(x) \Rightarrow Q(x, y) \\ \Delta_x = \{1\}, \Delta_y = \{1, 2, \dots, 10\} \end{matrix} \right) \right)^{10}$$

### Partitioning

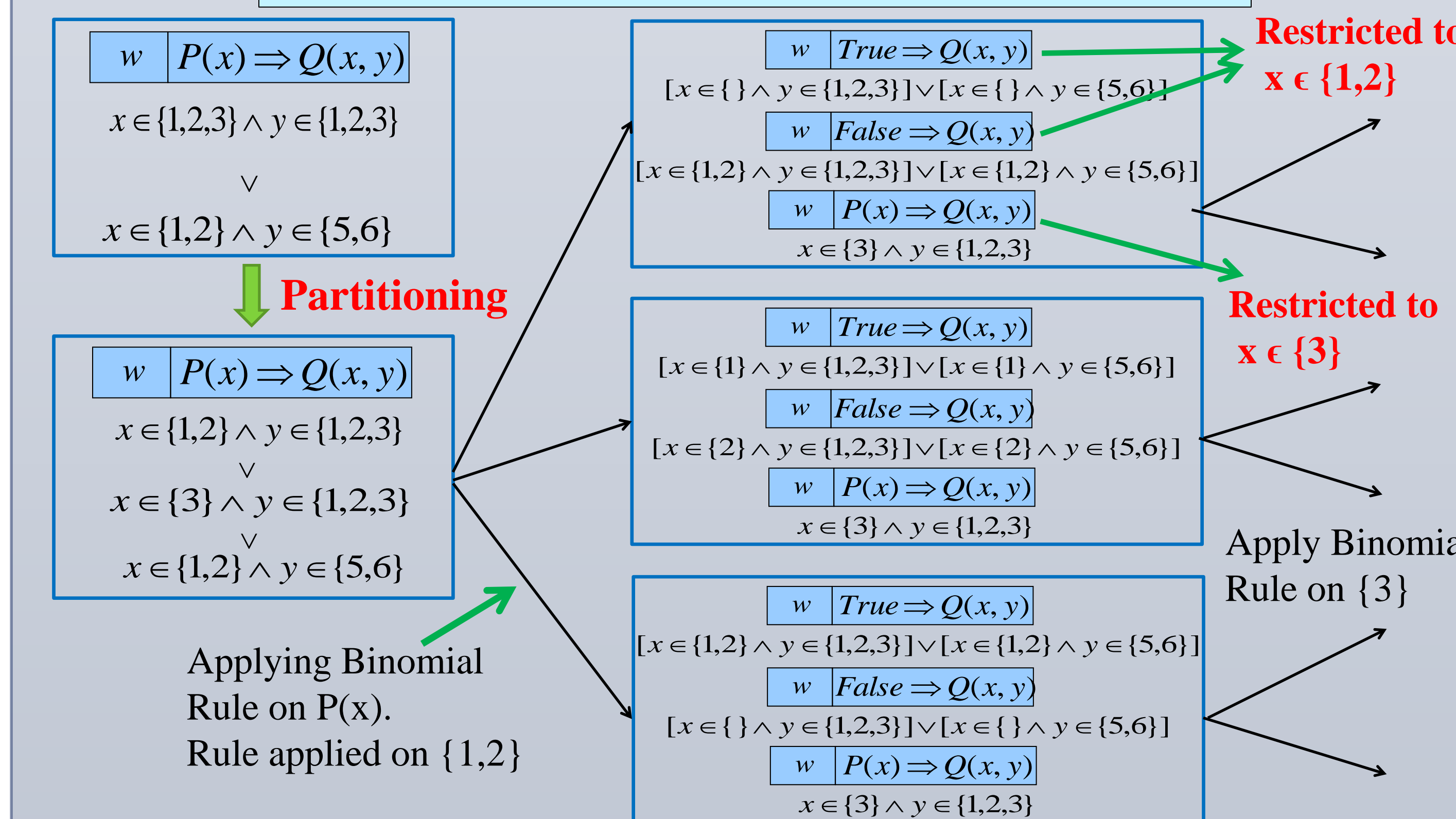
$$Z \left( \begin{matrix} w & P(x) \Rightarrow Q(x, y) \\ x \in \{1, 2\} \wedge y \in \{1, 2, 3\} \\ \vee \\ x \in \{3\} \wedge y \in \{1, 2, 3\} \\ \vee \\ x \in \{1, 2\} \wedge y \in \{5, 6\} \end{matrix} \right) = \left( Z \left( \begin{matrix} w & P(x) \Rightarrow Q(x, y) \\ x \in \{1\} \wedge y \in \{1, 2, 3\} \\ \vee \\ x \in \{1\} \wedge y \in \{5, 6\} \end{matrix} \right) \right)^2 * \left( Z \left( \begin{matrix} w & P(x) \Rightarrow Q(x, y) \\ x \in \{3\} \wedge y \in \{1, 2, 3\} \end{matrix} \right) \right)^1$$

## LIFTING RULE 2 : BINOMIAL RULE [Gogate & Domingos '11]

### Binomial Rule



### Binomial Rule With Constraints



## EVIDENCE PROCESSING & NORMAL FORM

**Normal Form** : Unconstrained representation in which

- No constant in any formula
- Variables appearing in same predicate position have identical domains..

Normal

$$w \quad P_1(x) \Rightarrow Q(y)$$

$$\Delta_x = \{1\}, \Delta_y = \{1, 2, \dots, n\}$$

...

$$w \quad P_{n/2}(x) \Rightarrow Q(y)$$

$$\Delta_x = \{n/2\}, \Delta_y = \{1, 2, \dots, n\}$$

...

$$w \quad P'(x) \Rightarrow Q(y)$$

$$\Delta_x = \{n/2+1, \dots, n\}, \Delta_y = \{1, 2, \dots, n\}$$

Evidence :

$$P(1) = P(2) = \dots = P(n/2) = \text{True}$$

$$x \in \{1, 2, \dots, n/2\} \wedge y \in \{1, 2, \dots, n\}$$

$$x \in \{n/2+1, \dots, n\} \wedge y \in \{1, 2, \dots, n\}$$

## PREVIOUS APPROACHES

Approach	Constraint Type	Constraint Aggregation	Tractable Solver	Lifting Algorithm
Lifted VE [Poole 2003]	eq/ineq no subset	intersection no union	No	Lifted VE
CFOVE [Milch et al 2008]	eq/ineq no subset	intersection no union	Yes	Lifted VE
GCFOVE [Taghipour et al 2012]	Subset (tree based) No inequality	Intersection Union	Yes	Lifted VE
Approx LBP [Singla et al 2014]	Subset (hypercube) No inequality	Intersection Union	No	Lifted message passing
Knowledge Compilation (KC) [Broeck et al 2011]	eq/ineq subset	Intersection no union	No	First order knowledge compilation
Lifted Inference from other side [Jha et al 2010]	Normal forms (no constraints)	None	Yes	Lifting rules : decomposer, binomial etc
PTP [Gogate & Domingos 2011]	eq/ineq no subset	Intersection no union	no	lifted search & sampling: Decomposer, binomial
Current Work	eq/ineq subset	Intersection Union	Yes	lifted search & sampling: Decomposer, binomial, single occurrence

- SetInEq is one of the most expressive constraint languages..
- SetInEq allows for efficient constraint processing during lifting.

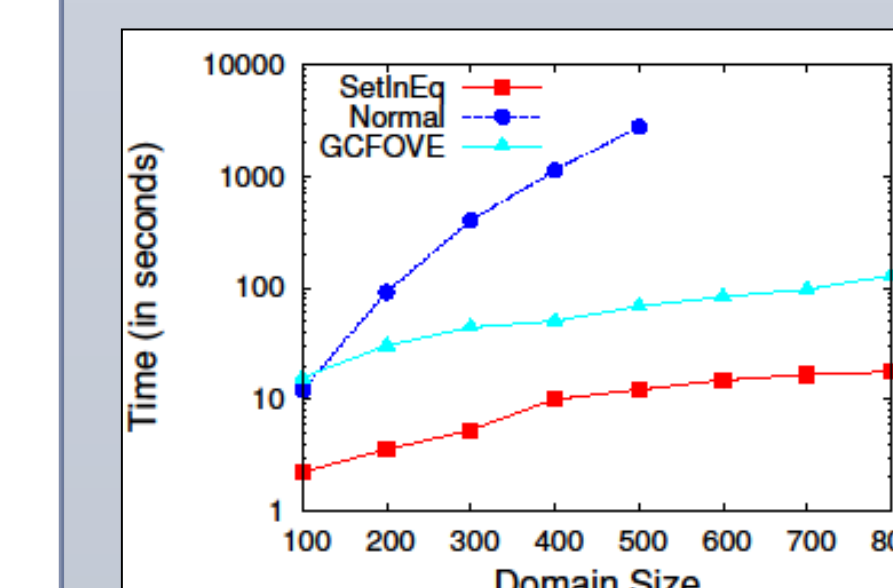
## EXPERIMENTS & RESULTS

- Algorithms Compared :

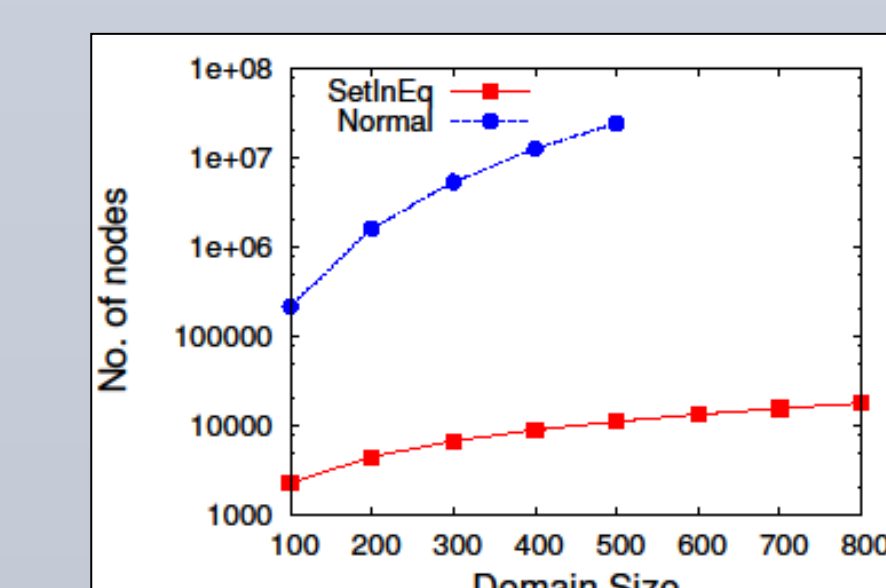
- SetInEq, Normal, GCFOVE, PTP

- Exact Inference

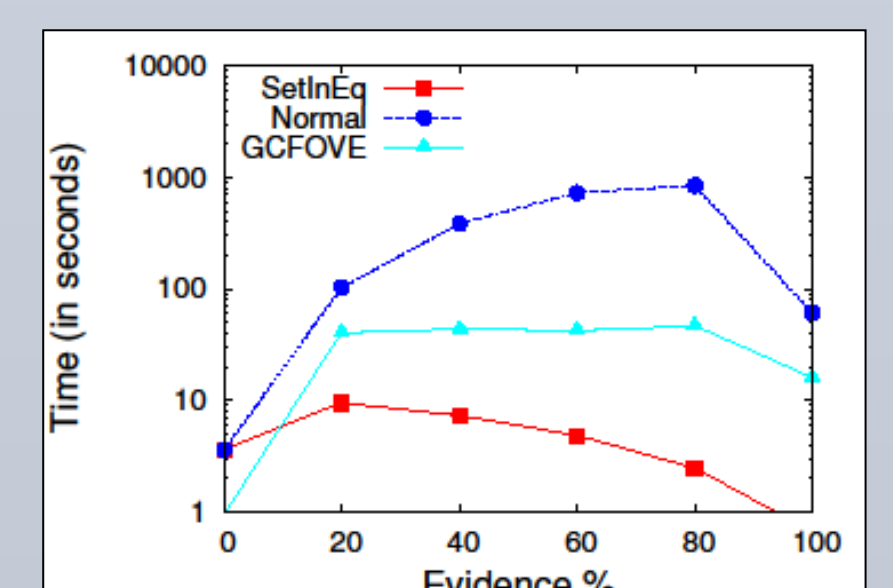
- Performed exact inference on Friends and Smokers (FS) and Professors and students (PS) datasets.
- PTP failed to scale to domain size of even 100.



FS : size vs time (sec)



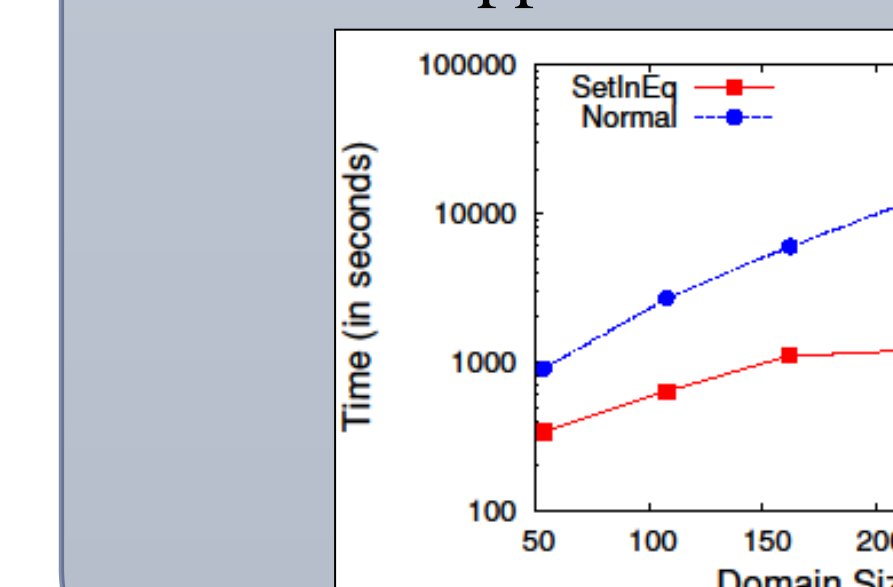
FS : size vs #nodes



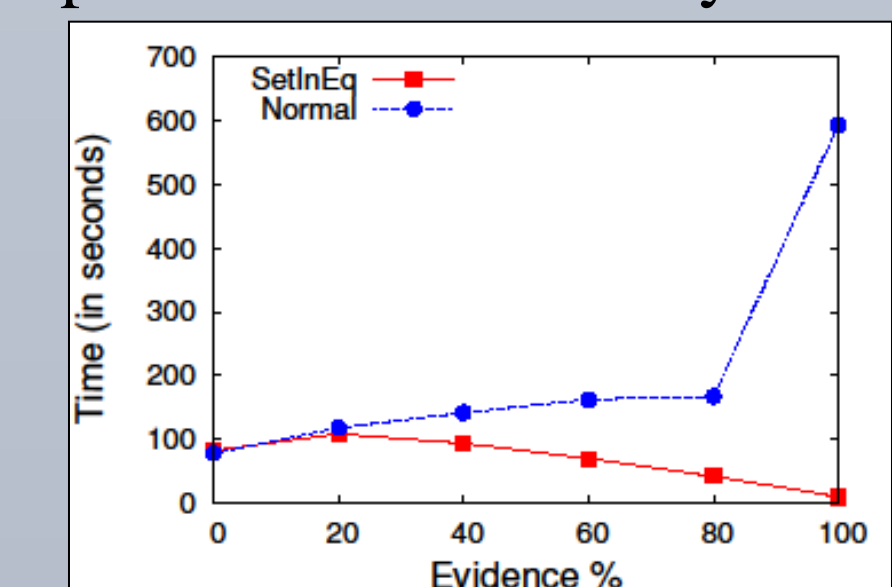
FS : evid% vs time (sec)

- Approximate Inference

- Performed approximate inference on WebKB and IMDB datasets
- GCFOVE doesn't have approximate version.
- Approximate version of PTP is not fully implemented in Alchemy2



WebKB : size vs time (sec)



IMDB : evid% vs time (sec)