

# Exploring Unknown Universes in Probabilistic Relational Models

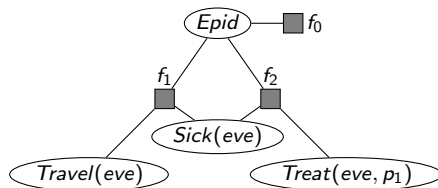
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December 4, 2019

# Probabilistic Graphical Models

Factor Graph  $F$ : **Semantics**

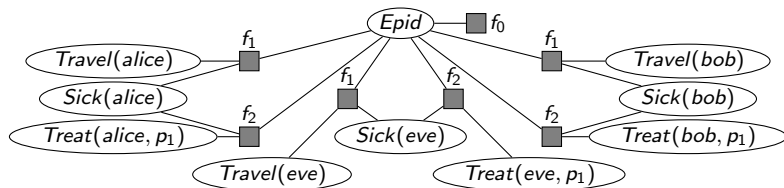


Compact encoding of full joint distribution

$$P_F = \frac{1}{Z} \prod_i f_i$$

# Probabilistic Graphical Models

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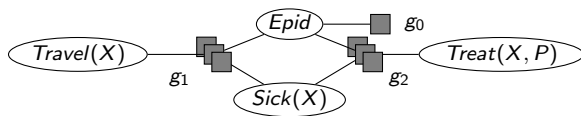


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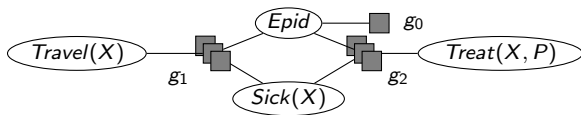
# Probabilistic Relational and Lifted Models

Parfactor Graph  $G$ : **Distribution Semantics** (Sato 1995)



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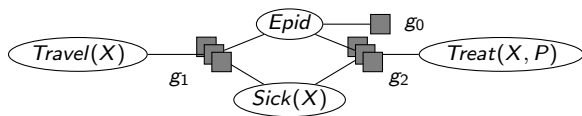


Compact encoding of full joint distribution  
with tractable inference in domain sizes

$$P_G = \frac{1}{Z} \prod_{f \in gr(G)} f$$

# Probabilistic Relational and Lifted Models

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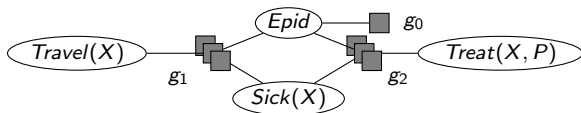
$$P_G = \frac{1}{Z} \prod_{f \in gr(G)} f$$



Requires knowing **domains** of parameters  $(X, P)$  in  $g_i$ 's  
(i.e., the universe is known)

# Known Universe

Parfactor Graph  $G$ : **Distribution Semantics** (e.g., Poole 2003, Taghipour 2013)



## Universe

General domains of parameters, e.g.,

$$dom(X) = \{alice, eve, bob\}$$

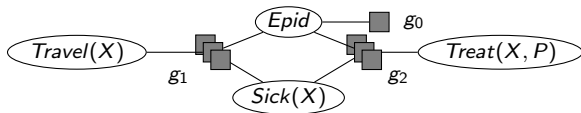
Constraint  $C_i$  for  $g_i$  as (subsets of) combinations of domains, e.g.,

$$C_2 = ((X, P), dom(X) \times dom(P))$$

Distribution semantics apply given a universe

# Problem: Unknown Universe

Parfactor Graph G: **Distribution Semantics**



## Ramifications

General domains of parameters are **empty**,

$$dom(X) = \{alice, eve, bob\}$$

Constraint  $C_i$  for  $g_i$  as combinations of domains **empty**

$$C_2 = ((X, P), \cancel{dom(X)} \times \cancel{dom(P)})$$

Semantics do not apply  $\rightarrow$  Lifted algorithms do not work



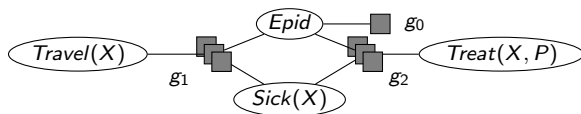
# Conference Contribution

## Semantics for Lifted Models with Unknown Universes

- Template model + constraint program
  - Local distributions as templates
  - Abstract description of constraints
- Domain worlds
  - Describe possible universes
  - Inputs to constraint program
- Generate a set of possible models
  - Within universe: distribution semantics apply
  - Reasoning over possible models

**Enable tractable inference with lifted algorithms again**

# Template Models



## Template Model

Local distributions as templates with empty constraints, e.g.,

$$C_2 = ((X, P), \perp)$$

# Template Models



## Template Model

Local distributions as templates with empty constraints, e.g.,

$$C_2 = ((X, P), \perp)$$

What do we do about empty constraints?

# Constraint Programs



## Constraint Program

Generate specific constraints for template models if given domain, e.g., using probabilistic Datalog:

```

element_of_C2(X,Y1) :- linked(X,Y1,Y2).
element_of_C2(X,Y2) :- linked(X,Y1,Y2).
linked(X,Y1,Y2) :- instance_of_X(X) & pair(Y1,Y2).
0.7 pair(t1,t2). 0.2 pair(t2,t3). 0.1 pair(t1,t3).
  
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# Constraint Programs



## Constraint Program

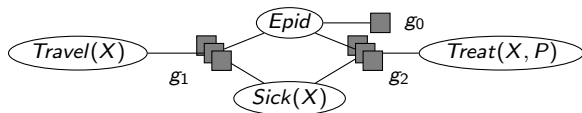
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Where do we get a given domain from?

# Domain Worlds



## Domain World

Specify or generate possible domains, e.g., for  $X$



# Distribution-based Semantics

## Semantics for Lifted Models with Unknown Universes

- Template model:
  - Empty constraints
- Constraint program:
  - Generate constraints given a domain world
- Domain worlds:
  - Generate possible domain worlds

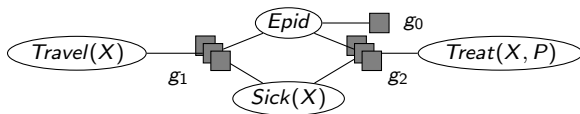
**Generate a set/distribution of possible models**



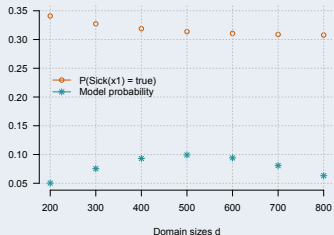
Distribution semantics apply again

# New Queries Emerging

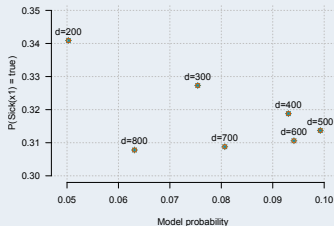
## Exploration



## Model and query probabilities



with respect to domain size



for skyline query

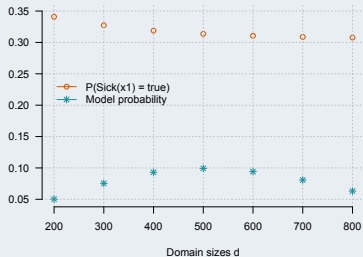


# New Queries Emerging

## Model Checking



## Behaviour between models



E.g., does the probability of

- an individual being sick decrease with larger domains?
- an epidemic decrease with more treatments?
- an epidemic rise if more people travel?

# Analysis: Probabilistic Inference with Unknown Universes

## Syntactic components

- Template model
- Constraint program
- Domain worlds

Set of possible worlds

- Distribution: Expected values
- Runtime efforts increase
- Possibilities for new queries arise

Constraint meta-programming

- Build oracle for algorithms

Transfer learning

- Decoupling from specific domain

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**Enable tractable inference with lifted algorithms again**