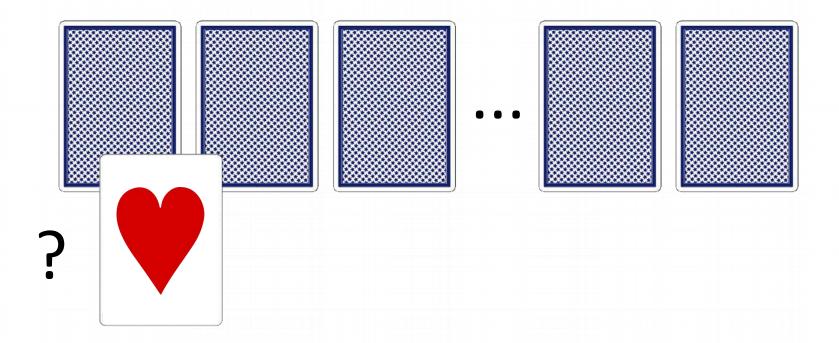
# First-Order Knowledge Compilation for Probabilistic Reasoning

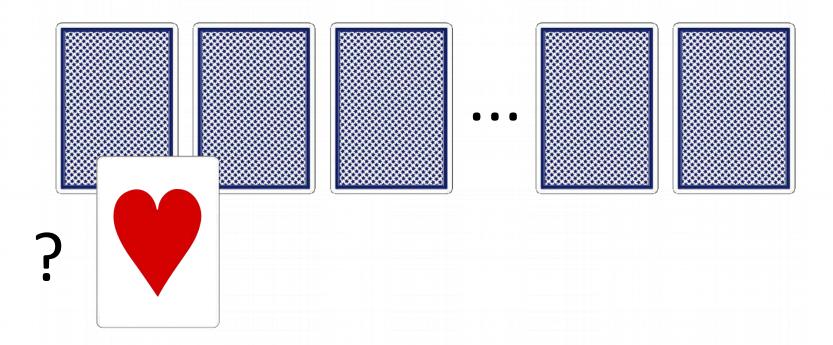
Guy Van den Broeck

based on joint work with Adnan Darwiche, Dan Suciu, and many others

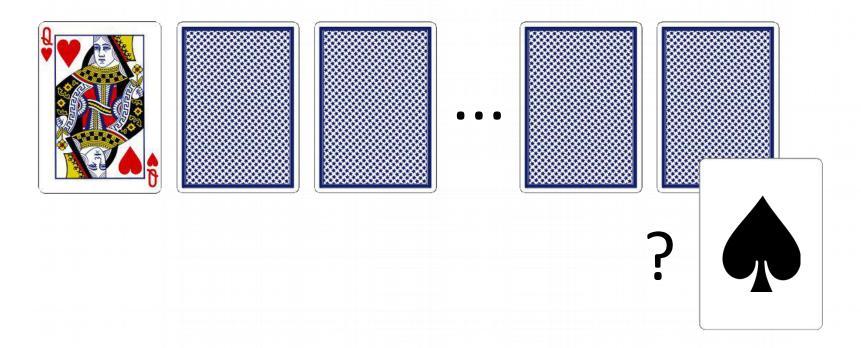
### **MOTIVATION 1**



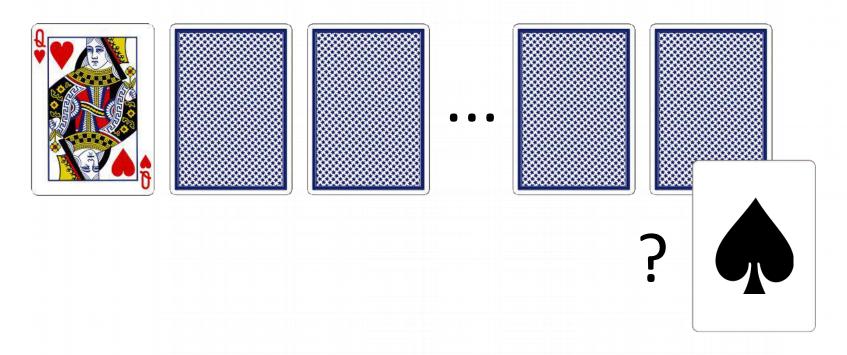
Probability that Card1 is Hearts?



Probability that Card1 is Hearts? 1/4



Probability that Card52 is Spades given that Card1 is QH?



Probability that Card52 is Spades given that Card1 is QH?

13/51

#### Let us automate this:

- 1. CNF encoding for deck of cards
- 2. Compile to tractable knowledge base (e.g., d-DNNF)
- 3. Condition on observations/questions "Card1 is hearts"
- 4. Model counting

#### Let us automate this:

- 1. CNF encoding for deck of cards
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A typical BeyondNP pipeline!

#### Let us automate this:

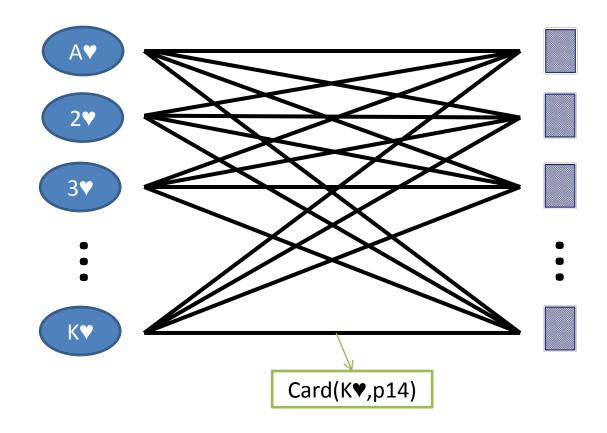
#### 1. CNF encoding for deck of cards

```
Card(p1,c1) v Card(p1,c2) v ...
Card(p1,c1) v Card(p2,c1) v ...
¬Card(p1,c1) v ¬Card(p1,c2)
¬Card(p1,c2) v ¬Card(p1,c3)
...
¬Card(p2,c1) v ¬Card(p2,c2)
...
```

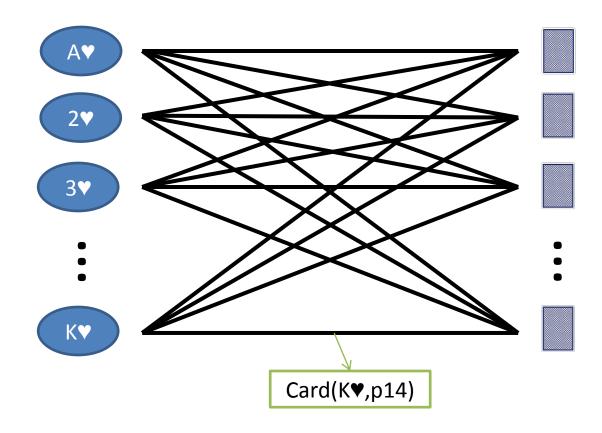
#### Let us automate this:

- 1. CNF encoding for deck of cards
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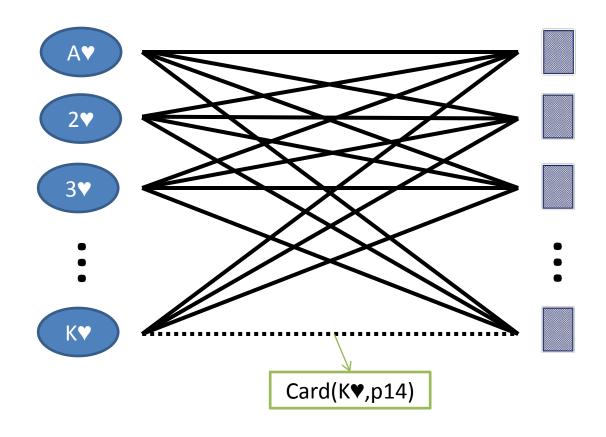
Which language to choose? Cards problem is easy: we want to be polynomial.



- 2. Compile to tractable knowledge base
- 3. Condition on observations/questions
- 4. Model counting

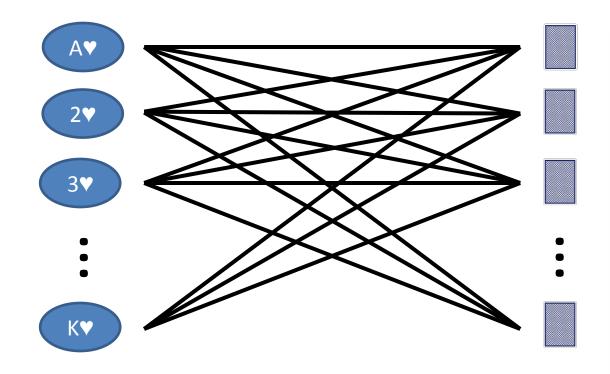


- 2. Compile to tractable knowledge base
- 3. Condition on observations/questions
- 4. Model counting

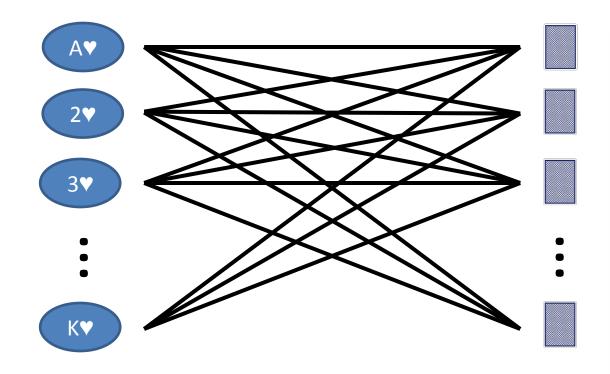


- 2. Compile to tractable knowledge base
- 3. Condition on observations/questions
- 4. Model counting

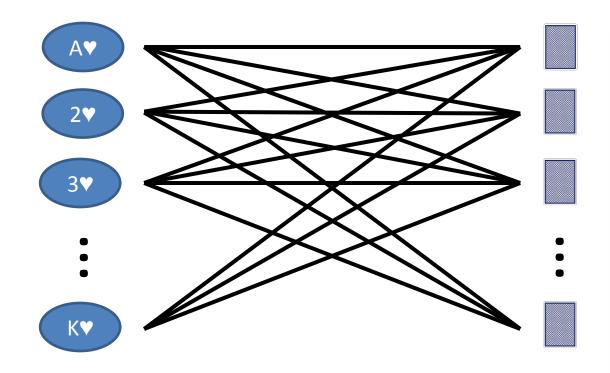
¬ Card(K♥,p14)



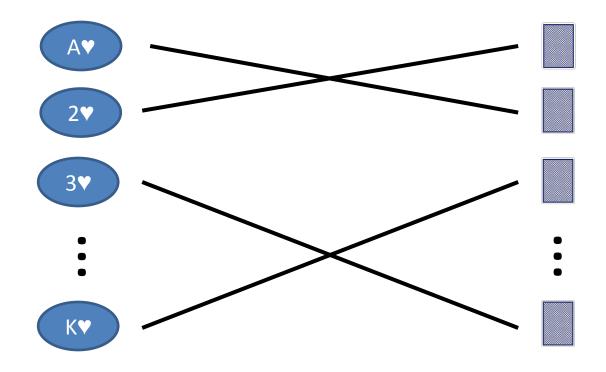
- 2. Compile to tractable knowledge base
- 3. Condition on observations/questions
- 4. Model counting



- 2. Compile to tractable knowledge base
- 3. Condition on observations/questions
- 4. Model counting



- 2. Compile to tractable knowledge base
- 3. Condition on observations/questions
- 4. Model counting: How many *perfect matchings*?

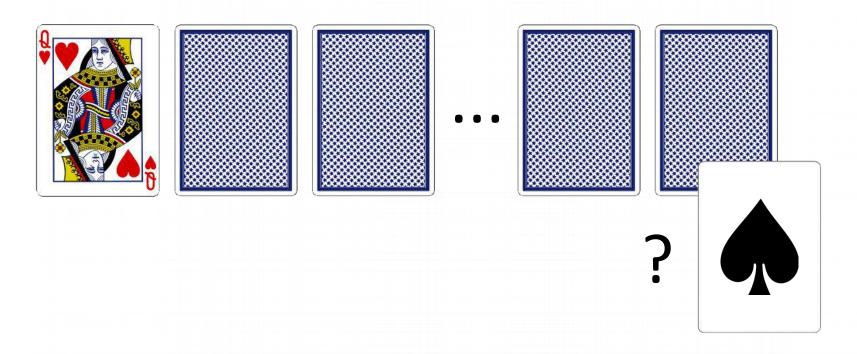


- 2. Compile to tractable knowledge base
- 3. Condition on observations/questions
- 4. Model counting: How many perfect matchings?

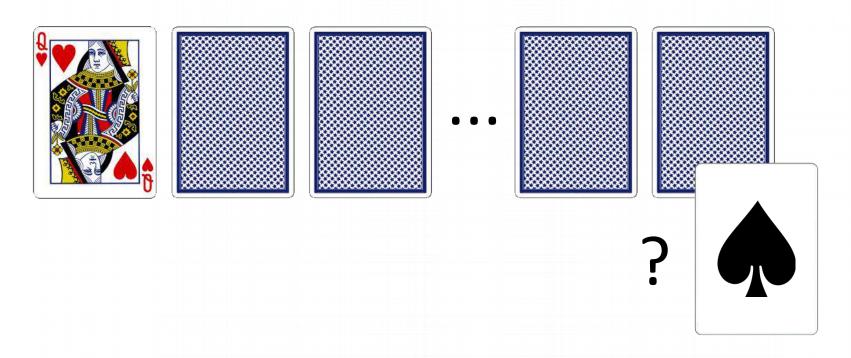
#### Observations

- Deck of cards = complete bigraph
- CD = removing edges in bigraph
   Encode any bigraph in cards problem
- CT = counting perfect matchings
- Problem is #P-complete!

No language with CD and CT can represent the cards problem compactly, unless P=NP.

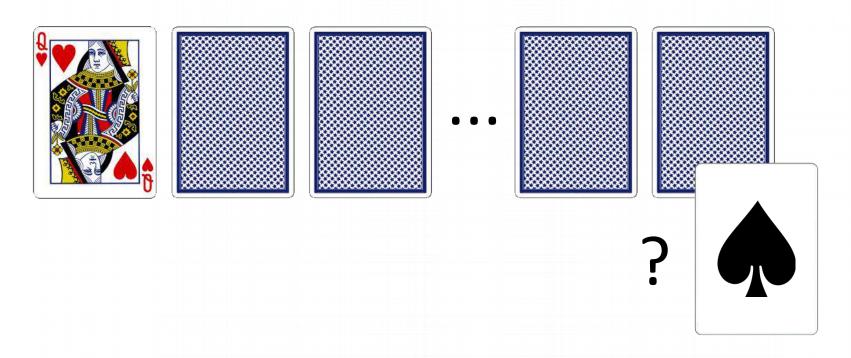


Probability that Card52 is Spades given that Card1 is QH?



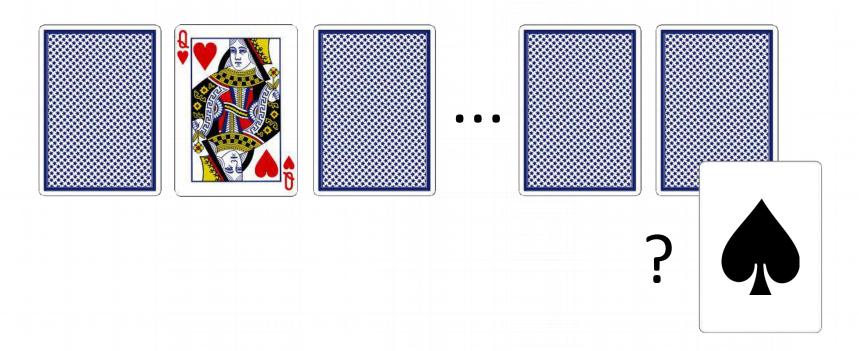
Probability that Card52 is Spades given that Card1 is QH?

13/51

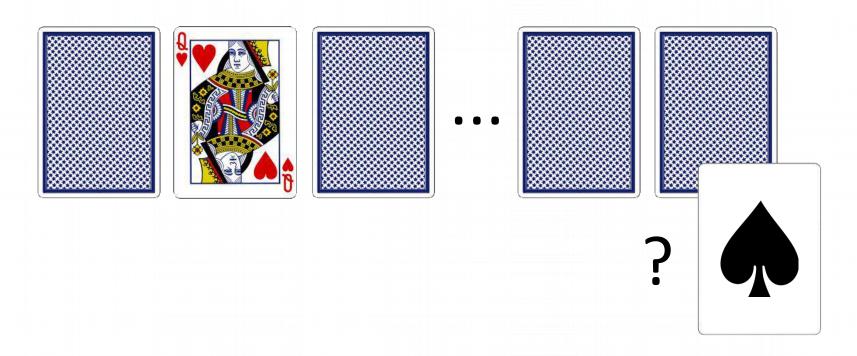


Probability that Card52 is Spades given that Card1 is QH?

13/51

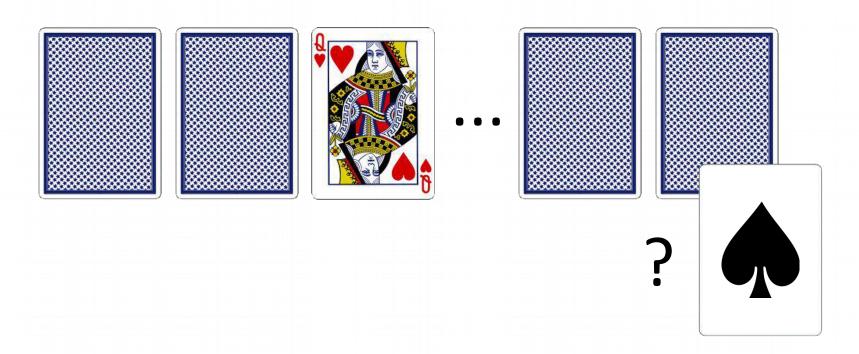


Probability that Card52 is Spades given that Card2 is QH?

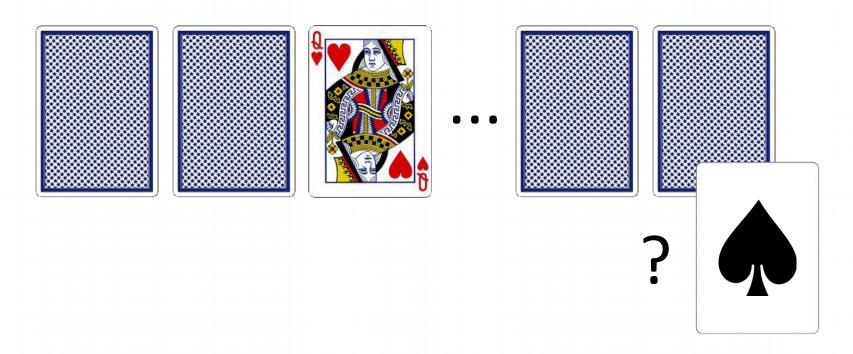


Probability that Card52 is Spades given that Card2 is QH?

13/51



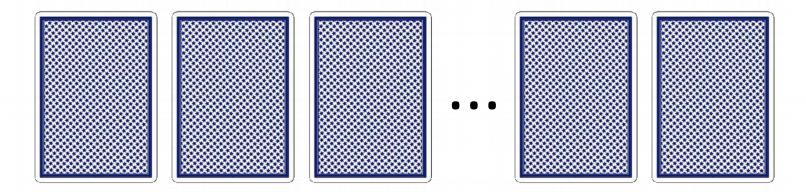
Probability that Card52 is Spades given that Card3 is QH?



Probability that Card52 is Spades given that Card3 is QH?

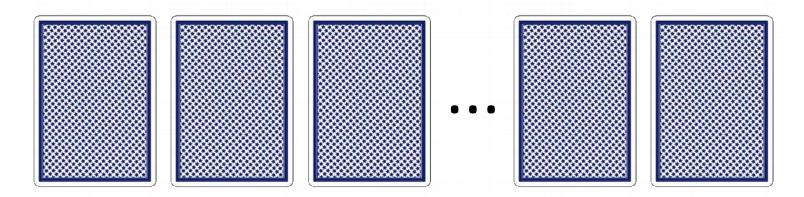
13/51

### Tractable Reasoning



What's going on here?
Which property makes reasoning tractable?

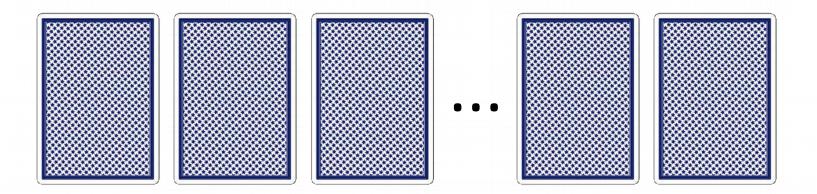
## Tractable Reasoning



What's going on here?
Which property makes reasoning tractable?

- High-level (first-order) reasoning
- Symmetry
- Exchangeability

**⇒ Lifted Inference** 



#### Let us automate this:

Relational/FO model

```
\forall p, \exists c, Card(p,c)

\forall c, \exists p, Card(p,c)

\forall p, \forall c, \forall c', Card(p,c) \land Card(p,c') \Rightarrow c = c'
```

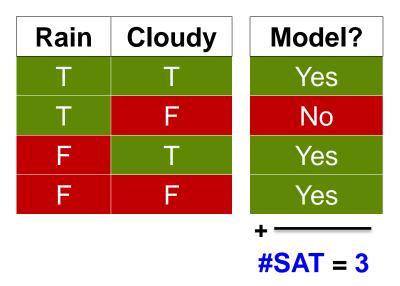
First-Order Knowledge Compilation

### MOTIVATION 2

### Model Counting

- Model = solution to a propositional logic formula Δ
- Model counting = #SAT

$$\triangle$$
 = (Rain  $\Rightarrow$  Cloudy)

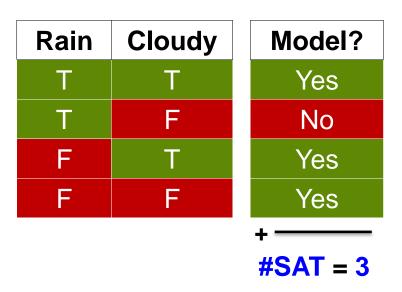


[Valiant] #P-hard, even for 2CNF

# Weighted Model Counting

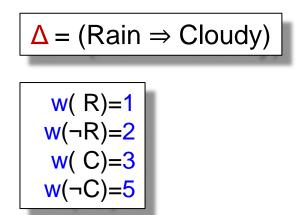
- Model = solution to a propositional logic formula Δ
- Model counting = #SAT

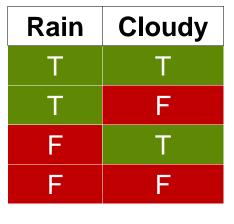
 $\Delta = (Rain \Rightarrow Cloudy)$ 



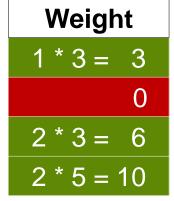
# Weighted Model Counting

- Model = solution to a propositional logic formula Δ
- Model counting = #SAT
- Weighted model counting (WMC)
  - Weights for assignments to variables
  - Model weight is product of variable weights w(.)



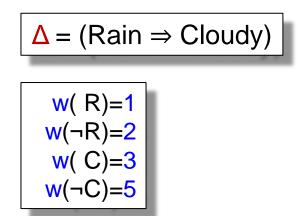


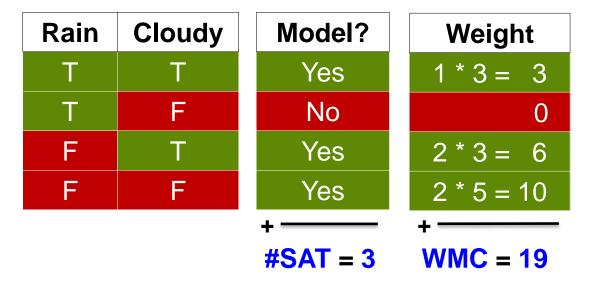




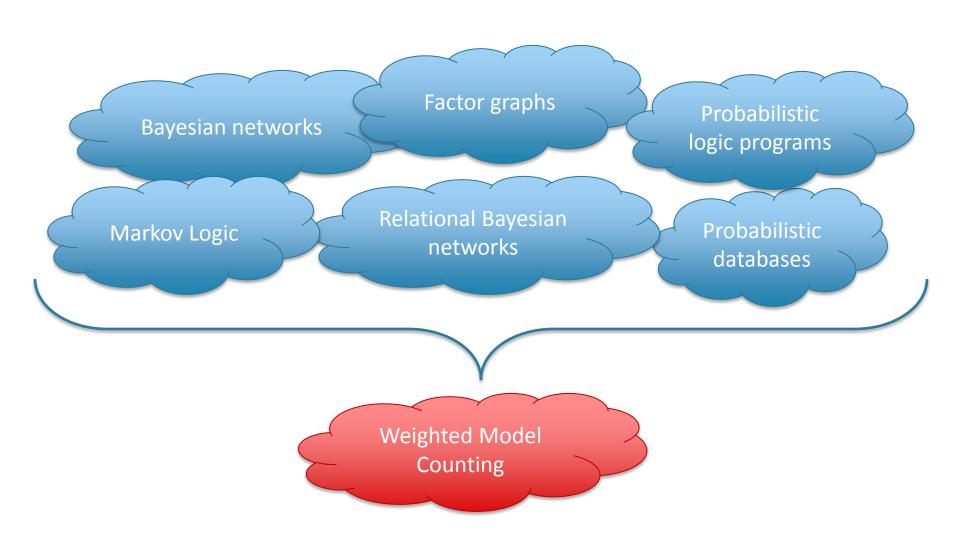
# Weighted Model Counting

- Model = solution to a propositional logic formula Δ
- Model counting = #SAT
- Weighted model counting (WMC)
  - Weights for assignments to variables
  - Model weight is product of variable weights w(.)





# Assembly language for probabilistic reasoning and learning



### First-Order Model Counting

Model = solution to first-order logic formula  $\Delta$ 

```
∆ = ∀d (Rain(d)

⇒ Cloudy(d))
```

Days = {Monday}

### First-Order Model Counting

Model = solution to first-order logic formula  $\Delta$ 



Days = {Monday}

Rain(M)	Cloudy(M)	Model?
Т	Т	Yes
Т	F	No
F	Т	Yes
F	F	Yes
		+

FOMC = 3

Model = solution to first-order logic formula  $\Delta$ 

 $\Delta$  = ∀d (Rain(d) ⇒ Cloudy(d))

Days = {Monday **Tuesday**}

Rain(M)	Cloudy(M)	Rain(T)	Cloudy(T)	Model?
Т	Т	Т	Т	Yes
Т	F	Т	Т	No
F	Т	Т	Т	Yes
F	F	Т	Т	Yes
Т	Т	Т	F	No
Т	F	Т	F	No
F	Т	Т	F	No
F	F	Т	F	No
Т	Т	F	Т	Yes
Ţ	F	F	T	No
F	Т	F	Т	Yes
F	F	F	Т	Yes
Т	Т	F	F	Yes
Т	F	F	F	No
F	Т	F	F	Yes
F	F	F	F	Yes

Model = solution to first-order logic formula  $\Delta$ 

 $\Delta$  = ∀d (Rain(d) ⇒ Cloudy(d))

Days = {Monday **Tuesday**}

Rain(M)	Cloudy(M)	Rain(T)	Cloudy(T)	Model?
Т	Т	Т	Т	Yes
Т	F	Т	Т	No
F	Т	Т	Т	Yes
F	F	Т	Т	Yes
Т	Т	Т	F	No
Т	F	Т	F	No
F	Т	Т	F	No
F	F	Т	F	No
Т	Т	F	Т	Yes
Т	F	F	Т	No
F	Т	F	Т	Yes
F	F	F	Т	Yes
Т	Т	F	F	Yes
Т	F	F	F	No
F	Т	F	F	Yes
F	F	F	F	Yes

Model = solution to first-order logic formula  $\Delta$ 

$$\Delta$$
 = ∀d (Rain(d)  
⇒ Cloudy(d))

Rain(M)	Cloudy(M)	Rain(T)	Cloudy(T)	Model?	Weight
Т	Т	Т	Т	Yes	1 * 1 * 3 * 3 = 9
Т	F	Т	Т	No	0
F	Т	Т	T	Yes	2 * 1* 3 * 3 = 18
F	F	Т	Т	Yes	2 * 1 * 5 * 3 = 30
Т	Т	Т	F	No	0
Т	F	Т	F	No	0
F	Т	Т	F	No	0
F	F	Т	F	No	0
Т	Т	F	Т	Yes	1 * 2 * 3 * 3 = 18
Т	F	F	Т	No	0
F	Т	F	Т	Yes	2 * 2 * 3 * 3 = 36
F	F	F	Т	Yes	2 * 2 * 5 * 3 = 60
Т	Т	F	F	Yes	1 * 2 * 3 * 5 = 30
Т	F	F	F	No	0
F	Т	F	F	Yes	2 * 2 * 3 * 5 = 60
F	F	F	F	Yes	2 * 2 * 5 * 5 = 100

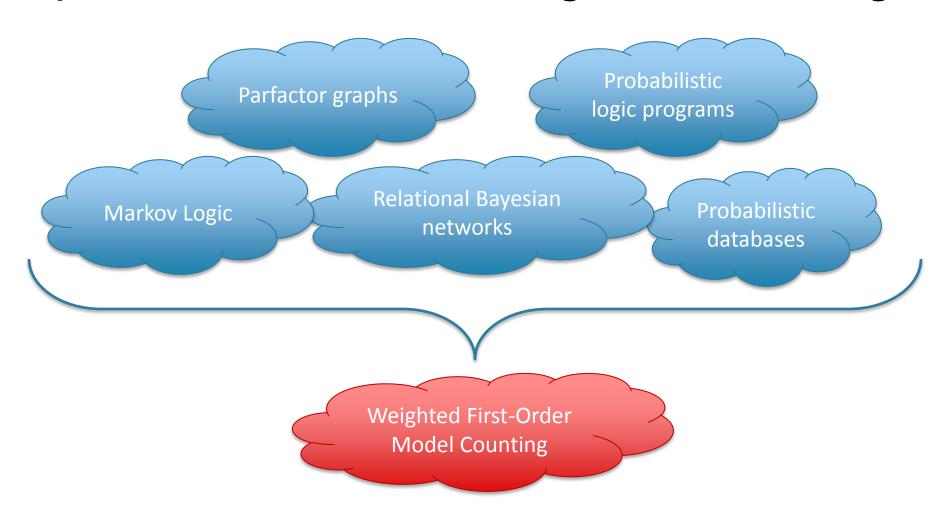
Model = solution to first-order logic formula  $\triangle$ 

```
\Delta = ∀d (Rain(d)

⇒ Cloudy(d))
```

Rain(M)	Cloudy(M)	Rain(T)	Cloudy(T)	Model?	Weight
Т	Т	Т	Т	Yes	1 * 1 * 3 * 3 = 9
Ţ	F	Т	T	No	0
F	Т	Т	Т	Yes	2 * 1* 3 * 3 = 18
F	F	Т	Т	Yes	2 * 1 * 5 * 3 = 30
Т	Т	Т	F	No	0
Т	F	Т	F	No	0
F	Т	Т	F	No	0
F	F	Т	F	No	0
Т	Т	F	Т	Yes	1 * 2 * 3 * 3 = 18
T	F	F	Т	No	0
F	Т	F	Т	Yes	2 * 2 * 3 * 3 = 36
F	F	F	Т	Yes	2 * 2 * 5 * 3 = 60
Т	Т	F	F	Yes	1 * 2 * 3 * 5 = 30
Т	F	F	F	No	0
F	Т	F	F	Yes	2 * 2 * 3 * 5 = 60
F	F	F	F	Yes	2 * 2 * 5 * 5 = 100

# Assembly language for high-level probabilistic reasoning and learning



# Statistical Relational Learning

```
Hard constraint

\infty
 Smoker(x) \Rightarrow Person(x)

Soft constraint

3.75
 Smoker(x) \land Friend(x,y) \Rightarrow Smoker(y)
```

- An MLN = set of constraints  $(\mathbf{w}, \Gamma(\mathbf{x}))$
- Weight of a world = product of w, for all rules (w, Γ(x)) and groundings Γ(a) that hold in the world

 $P_{MLN}(Q) = [sum of weights of worlds of Q] / Z$ 

Applications: large probabilistic KBs

## FO NNF SYNTAX

## First-Order Knowledge Compilation

- Input: Sentence in FOL
- Output: Representation tractable for some class of queries.
- In this work:
  - Function-free FOL
  - Model counting in NNF tradition
- Some pre-KC-map work:
  - FO Horn clauses
  - FO BDDs

# Alphabet

- FOL
  - Predicates/relations: Friends
  - Object names: x, y, z
  - Object variables: X, Y, Z
  - Symbols classical FOL (∀, ∃, ∧, ∨, ¬,...)
- Group logic
  - Group variables: X, Y, Z
  - Symbols from basic set theory(e.g., ∪, ∩, ∈, ⊆, {, }, complement).

# Syntax

- Object terms: X, alice, bob
- Group terms : X, {alice,bob}, X ∪ Y
- Atom: Friends(alice,X)
- Formulas:
  - $-(\alpha)$ ,  $\neg \alpha$ ,  $\alpha \vee \beta$ , and  $\alpha \wedge \beta$
  - $\forall X \in \mathbf{G}$ ,  $\alpha$  and  $\exists X \in \mathbf{G}$ ,  $\alpha$
  - $\forall X \subseteq G$ ,  $\alpha$  and  $\exists X \subseteq G$ ,  $\alpha$
- Group logic syntactic sugar:
  - P(G) is  $\forall X \in G, P(X)$
  - $-\overline{P}(G)$  is  $\forall X \in G, \neg P(X)$

# Examples:

∀X ∈ G, Y ∈ {alice, bob},
 Enemies(X, Y)
 ⇒¬Friends(X, Y) ∧ ¬Friends(Y, X)

•  $\forall X \in G, Y \in G,$ Smokes(X)  $\land$  Friends(X, Y)  $\Rightarrow$  Smokes(Y)

•  $\exists \mathbf{G} \subseteq \{\text{alice, bob}\}$ ,  $Smokes(\mathbf{G}) \land Healthy(\mathbf{G})$ 

#### **Semantics**

- Template language for propositional logic
- Grounding a sentence: gr(α)
  - Replace ∀ by ∧
  - Replace ∃ by ∨
  - End result: ground sentence = propositional logic
- Grounding is polynomial in group sizes
   when no ∀X ⊆ G or ∃X ⊆ G
   Important for polytime reduction to NNF circuits

# Decomposability

• Conjunction:  $\alpha(X,G) \wedge \beta(X,G)$ 

For any substitution X=c and G=g, we have that  $gr(\alpha(c,g)) \land gr(\beta(c,g))$  is decomposable

Meaning:  $\alpha$  and  $\beta$  can never talk about the same ground atoms

• Quantifier:  $\forall Y \in G$ ,  $\alpha(Y)$ 

For any two a,b  $\in$  **G**, we have that  $gr(\alpha(a)) \land gr(\alpha(b))$  is decomposable

#### Determinism

• Disjunction:  $\alpha(X,G) \vee \beta(X,G)$ 

For any substitution X=c and G=g, we have that  $gr(\alpha(c,g)) \vee gr(\beta(c,g))$  is deterministic

Meaning:  $\alpha \wedge \beta$  is UNSAT

• Quantifier:  $\exists Y \in G$ ,  $\alpha(Y)$ 

For any two a,b  $\in$  **G**, we have that  $gr(\alpha(a)) \vee gr(\alpha(b))$  is decomposable

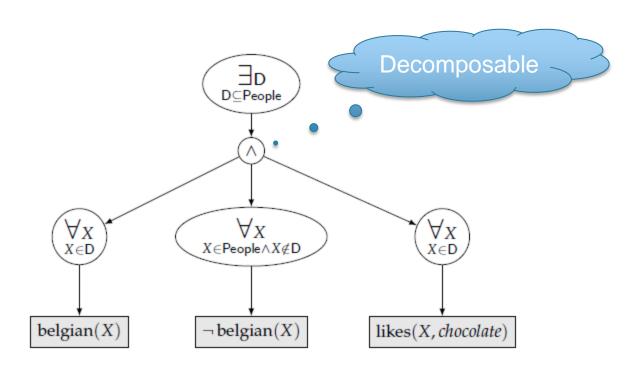
# **Group Quantifiers**

- Decomposability: ∀X ⊆ G, α(X)
   For any two A,B ⊆ G, we have that gr(α(A)) ∨ gr(α(B)) is decomposable
- Determinism: ∃X ⊆ G, α(X)
   For any two A,B ⊆ G, we have that gr(α(A)) ∨ gr(α(B)) is deterministic

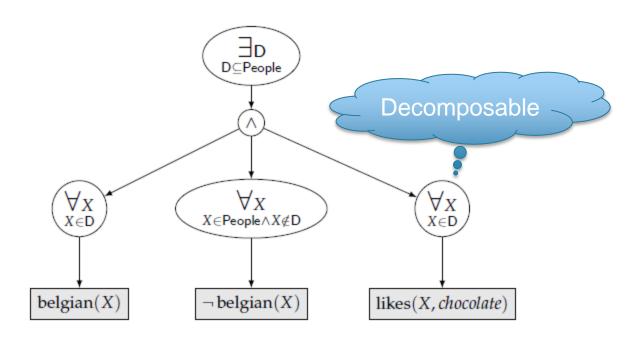
# Automorphism

- Object permutation σ : D→ D is a one-to-one mapping from objects to objects.
- Permuting  $\alpha$  using  $\sigma$  replaces  $\sigma$  in  $\sigma$  by  $\sigma(\sigma)$ .
- Sentences α and β are p-equivalent iff α is equivalent to an object permutation of β.
   Smokes(alice) and Smokes(bob) are p-equivalent
- Group quantifiers: ∀X ⊆ G, α(X) or ∃X ⊆ G, α(X)
   Are automorphic iff for any two A,B ⊆ G s.t.
   |A|=|B|, gr(α(A)) and gr(α(B)) are p-equivalent

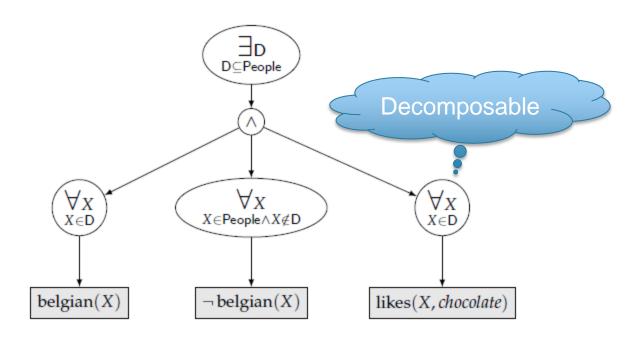
#### First-Order NNF



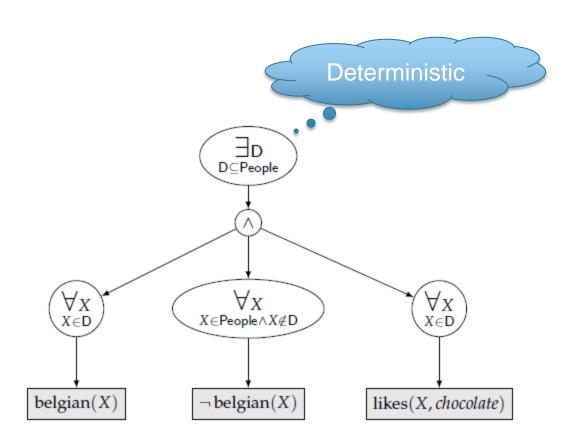
#### First-Order NNF



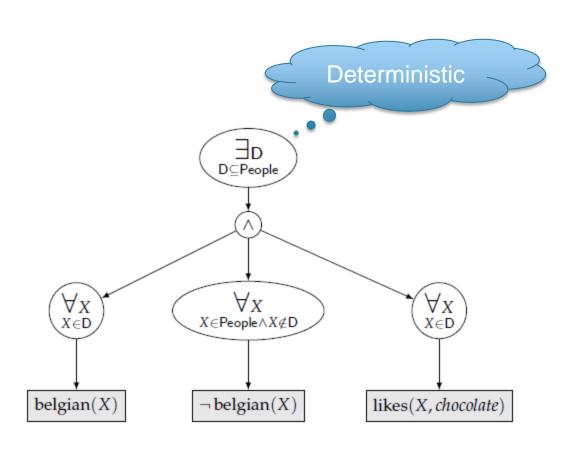
#### First-Order DNNF



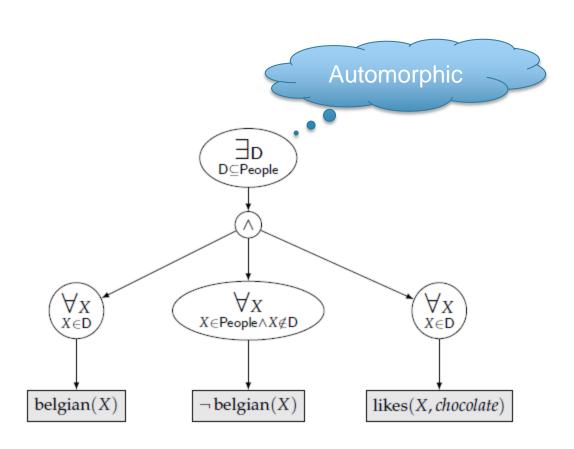
#### First-Order DNNF



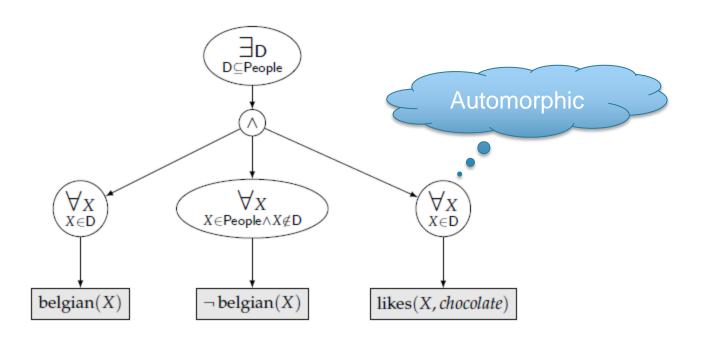
#### First-Order d-DNNF



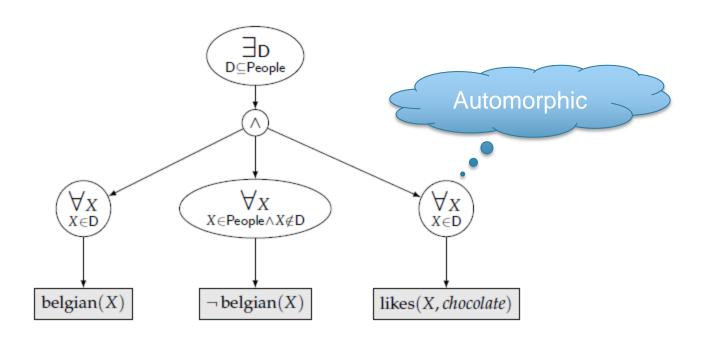
#### First-Order d-DNNF



#### First-Order d-DNNF



#### First-Order ad-DNNF



# FO NNF Languages

- FO NNF: group logic circuits, negation only on atoms
- FO d-DNNF: determinism and decomposability
   Grounding generates a d-DNNF
- FO DNNF
   Grounding generates a DNNF
- FO ad-DNNF: automorphic Powerful properties!

## FO NNF TRACTABILITY

# Symmetric WFOMC

**Def**. A weighted vocabulary is (R, w), where

```
-R = (R_1, R_2, ..., R_k) = relational vocabulary
```

- $w = (w_1, w_2, ..., w_k) = weights$
- Fix an FO formula Q, domain of size n
- The weight of a ground tuple t in R<sub>i</sub> is w<sub>i</sub>

```
Complexity of FOMC / WFOMC(Q, n)?

Data/domain complexity:

fixed Q, input n / and w
```

# Symmetric WFOMC on FO ad-DNNF

```
U(\alpha) = \begin{cases} 0 & \text{when } \alpha = \mathsf{false} \\ 1 & \text{when } \alpha = \mathsf{true} \\ 0.5 & \text{when } \alpha \text{ is a literal} \\ U(\ell_1) \times \dots \times U(\ell_n) & \text{when } \alpha = \ell_1 \wedge \dots \wedge \ell_n \\ U(\ell_1) + \dots + U(\ell_n) & \text{when } \alpha = \ell_1 \vee \dots \vee \ell_n \\ \prod_{i=1}^n U(\beta\{X/x_i\}) & \text{when } \alpha = \forall X \in \tau, \beta \text{ and } x_1, \dots, x_n \text{ are the objects in } \tau. \\ \sum_{i=1}^n U(\beta\{X/x_i\}) & \text{when } \alpha = \exists X \in \tau, \beta \text{ and } x_1, \dots, x_n \text{ are the objects in } \tau. \\ \prod_{i=0}^{|\tau|} U(\beta\{X/x_i\})^{\binom{|\tau|}{i}} & \text{when } \alpha = \exists X \in \tau, \beta, \text{ and } x_i \text{ is any subset of } \tau \text{ such that } |x_i| = i. \\ \sum_{i=0}^{|\tau|} \binom{|\tau|}{i} \cdot U(\beta\{X/x_i\}) & \text{when } \alpha = \exists X \subseteq \tau, \beta, \text{ and } x_i \text{ is any subset of } \tau \text{ such that } |x_i| = i. \end{cases}
```

Complexity polynomial in domain size! Polynomial in NNF size for bounded depth.

FO-Model Counting:  $w(R) = w(\neg R) = 1$ FO ad-DNNF sentences

FO-Model Counting:  $w(R) = w(\neg R) = 1$ FO ad-DNNF sentences

4. 
$$\Delta = (Stress(Alice) \Rightarrow Smokes(Alice))$$

Domain = {Alice}

FO-Model Counting:  $w(R) = w(\neg R) = 1$ FO ad-DNNF sentences

4. 
$$\triangle = (Stress(Alice) \Rightarrow Smokes(Alice))$$

 $\rightarrow$  3 models

Domain = {Alice}

FO-Model Counting: 
$$w(R) = w(\neg R) = 1$$
  
FO ad-DNNF sentences

4. 
$$\triangle = (Stress(Alice) \Rightarrow Smokes(Alice))$$

Domain = {Alice}

 $\rightarrow$  3 models

3. 
$$\triangle = \forall x$$
, (Stress(x)  $\Rightarrow$  Smokes(x))

Domain = {n people}

FO-Model Counting:  $w(R) = w(\neg R) = 1$ FO ad-DNNF sentences

4.  $\Delta = (Stress(Alice) \Rightarrow Smokes(Alice))$ 

Domain = {Alice}

 $\rightarrow$  3 models

3.  $\triangle = \forall x$ , (Stress(x)  $\Rightarrow$  Smokes(x))

Domain = {n people}

 $\rightarrow$  3<sup>n</sup> models

3.  $\triangle = \forall x$ , (Stress(x)  $\Rightarrow$  Smokes(x))

Domain = {n people}

 $\rightarrow$  3<sup>n</sup> models

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Domain = {n people}

 $\rightarrow$  3<sup>n</sup> models

2.  $\triangle = \forall y$ , (ParentOf(y)  $\land$  Female  $\Rightarrow$  MotherOf(y))

D = {n people}

3. 
$$\triangle = \forall x$$
, (Stress(x)  $\Rightarrow$  Smokes(x))

Domain = {n people}

 $\rightarrow$  3<sup>n</sup> models

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$$\triangle = \forall y$$
, (ParentOf(y)  $\land$  Female  $\Rightarrow$  MotherOf(y))

D = {n people}

$$\triangle$$
 =  $\forall y$ , (ParentOf(y)  $\Rightarrow$  MotherOf(y))

 $\rightarrow$  3<sup>n</sup> models

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$$\triangle = \forall x$$
, (Stress(x)  $\Rightarrow$  Smokes(x))

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 $\rightarrow$  3<sup>n</sup> models

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$$\triangle = \forall y$$
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D = {n people}

$$\triangle = \forall y, (ParentOf(y) \Rightarrow MotherOf(y))$$

$$\rightarrow$$
 3<sup>n</sup> models

$$\Delta$$
 = true

$$\rightarrow$$
 4<sup>n</sup> models

3. 
$$\Delta = \forall x$$
, (Stress(x)  $\Rightarrow$  Smokes(x))

Domain = {n people}

 $\rightarrow$  3<sup>n</sup> models

2. 
$$\triangle = \forall y$$
, (ParentOf(y)  $\land$  Female  $\Rightarrow$  MotherOf(y))

D = {n people}

$$\triangle = \forall y, (ParentOf(y) \Rightarrow MotherOf(y))$$

 $\rightarrow$  3<sup>n</sup> models

$$\Delta$$
 = true

 $\rightarrow$  4<sup>n</sup> models

$$\rightarrow$$
 3<sup>n</sup> + 4<sup>n</sup> models

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$$\triangle = \forall x$$
, (Stress(x)  $\Rightarrow$  Smokes(x))

Domain = {n people}

 $\rightarrow$  3<sup>n</sup> models

2. 
$$\triangle = \forall y$$
, (ParentOf(y)  $\land$  Female  $\Rightarrow$  MotherOf(y))

D = {n people}

$$\triangle = \forall y$$
, (ParentOf(y)  $\Rightarrow$  MotherOf(y))

 $\rightarrow$  3<sup>n</sup> models

 $\rightarrow$  4<sup>n</sup> models

$$\Delta$$
 = true

 $\rightarrow$  3<sup>n</sup> + 4<sup>n</sup> models

1. 
$$\Delta = \forall x, \forall y, (ParentOf(x,y) \land Female(x) \Rightarrow MotherOf(x,y))$$

D = {n people}

3. 
$$\triangle = \forall x$$
, (Stress(x)  $\Rightarrow$  Smokes(x))

Domain = {n people}

$$\rightarrow$$
 3<sup>n</sup> models

2. 
$$\triangle = \forall y$$
, (ParentOf(y)  $\land$  Female  $\Rightarrow$  MotherOf(y))

D = {n people}

$$\triangle = \forall y, (ParentOf(y) \Rightarrow MotherOf(y))$$

 $\rightarrow$  3<sup>n</sup> models

 $\rightarrow$  4<sup>n</sup> models

$$\Delta$$
 = true

 $\rightarrow$  3<sup>n</sup> + 4<sup>n</sup> models

1. 
$$\Delta = \forall x, \forall y, (ParentOf(x,y) \land Female(x) \Rightarrow MotherOf(x,y))$$

D = {n people}

$$\rightarrow$$
 (3<sup>n</sup> + 4<sup>n</sup>)<sup>n</sup> models

```
\Delta = \forall x, y \in \mathbf{D}, (Smokes(x) \land Friends(x,y) \Rightarrow Smokes(y))
```

Domain = {n people}

- Not decomposable!
- Rewrite as FO ad-DNNF:

```
\exists \mathbf{G} \subseteq \mathbf{D}, Smokes(\mathbf{G}) \land \overline{\mathsf{S}}mokes(\overline{\mathbf{G}}) \land \overline{\mathsf{F}}riends(\mathbf{G}, \overline{\mathbf{G}})
```

- Not possible to ground to d-DNNF!
- How to do tractable CT?

```
\sum_{i=0}^{|\tau|} {|\tau| \choose i} \cdot U(\beta\{\mathbf{X}/\mathbf{x}_i\}) \quad \text{when } \alpha = \exists \mathbf{X} \subseteq \tau, \beta, \text{ and } \mathbf{x}_i \text{ is any subset of } \tau \text{ such that } |\mathbf{x}_i| = i
```

 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

• If we know **G** precisely: who smokes, and there are *k* smokers?

#### **Database:**

Smokes(Alice) = 1 Smokes(Bob) = 0 Smokes(Charlie) = 0 Smokes(Dave) = 1 Smokes(Eve) = 0

k n-k

**Smokes** 

Friends Smokes

k

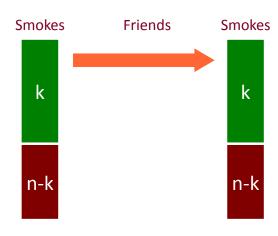
n-k

 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

• If we know **G** precisely: who smokes, and there are *k* smokers?

#### **Database:**

Smokes(Alice) = 1 Smokes(Bob) = 0 Smokes(Charlie) = 0 Smokes(Dave) = 1 Smokes(Eve) = 0

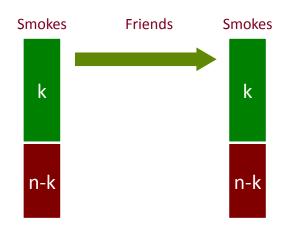


 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

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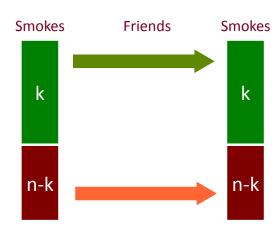


 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

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Smokes(Alice) = 1 Smokes(Bob) = 0 Smokes(Charlie) = 0 Smokes(Dave) = 1 Smokes(Eve) = 0

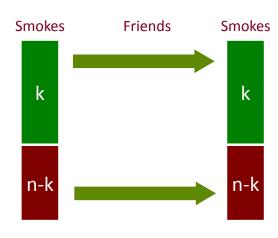


 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

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#### **Database:**

Smokes(Alice) = 1 Smokes(Bob) = 0 Smokes(Charlie) = 0 Smokes(Dave) = 1 Smokes(Eve) = 0



 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

• If we know **G** precisely: who smokes, and there are *k* smokers?

#### **Database:**

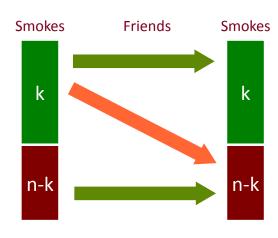
Smokes(Alice) = 1

Smokes(Bob) = 0

Smokes(Charlie) = 0

Smokes(Dave) = 1

Smokes(Eve) = 0



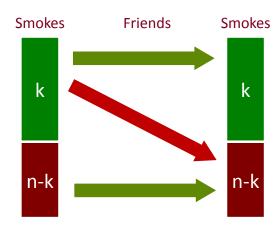
 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

• If we know **G** precisely: who smokes, and there are *k* smokers?

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Smokes(Alice) = 1 Smokes(Bob) = 0 Smokes(Charlie) = 0 Smokes(Dave) = 1

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 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

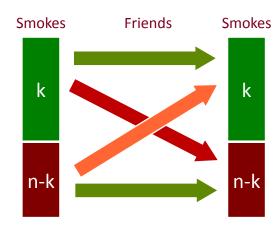
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Smokes(Dave) = 1

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 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

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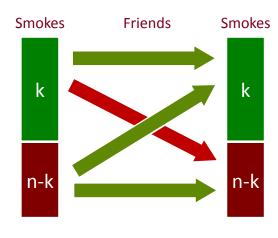
#### **Database:**

Smokes(Alice) = 1 Smokes(Bob) = 0

Smokes(Charlie) = 0

Smokes(Dave) = 1

Smokes(Eve) = 0



 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

• If we know **G** precisely: who smokes, and there are *k* smokers?

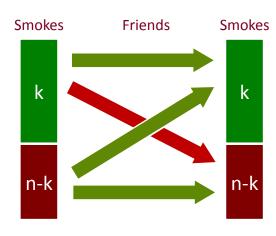
#### **Database:**

Smokes(Alice) = 1 Smokes(Bob) = 0 Smokes(Charlie) = 0

Smokes(Dave) = 1

Smokes(Eve) = 0

$$\rightarrow 2^{n^2-k(n-k)}$$
 models

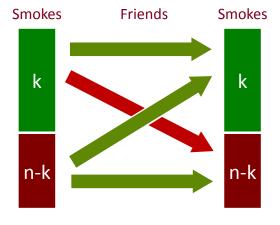


$$\exists \mathbf{G} \subseteq \mathbf{D}$$
, Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

If we know G precisely: who smokes, and there are k smokers?

#### **Database:**

Smokes(Alice) = 1 Smokes(Bob) = 0 Smokes(Charlie) = 0 Smokes(Dave) = 1 Smokes(Eve) = 0 ...  $\Rightarrow 2^{n^2-k(n-k)} \text{ models}$ 



• If we know that there are k smokers?

 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

If we know G precisely: who smokes, and there are k smokers?

#### **Database:**

Smokes(Alice) = 1 Smokes(Bob) = 0

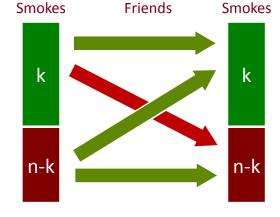
Smokes(Charlie) = 0

Smokes(Dave) = 1

Smokes(Eve) = 0

...

$$\rightarrow 2^{n^2-k(n-k)}$$
 models



• If we know that there are *k* smokers?

$$\rightarrow \binom{n}{k} 2^{n^2 - k(n-k)}$$
 models

 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

If we know G precisely: who smokes, and there are k smokers?

#### **Database:**

Smokes(Alice) = 1

Smokes(Bob) = 0

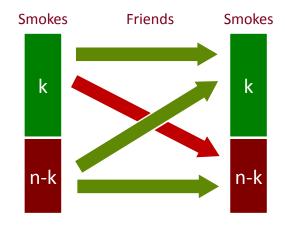
Smokes(Charlie) = 0

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Smokes(Eve) = 0

...

$$\rightarrow 2^{n^2-k(n-k)}$$
 models



• If we know that there are *k* smokers?

$$\rightarrow \binom{n}{k} 2^{n^2 - k(n-k)}$$
 models

In total...

 $\exists \mathbf{G} \subseteq \mathbf{D}$ , Smokes( $\mathbf{G}$ )  $\land \overline{\mathsf{S}}$ mokes( $\overline{\mathbf{G}}$ )  $\land \overline{\mathsf{F}}$ riends( $\mathbf{G}$ ,  $\overline{\mathbf{G}}$ )

If we know G precisely: who smokes, and there are k smokers?

#### **Database:**

Smokes(Alice) = 1

Smokes(Bob) = 0

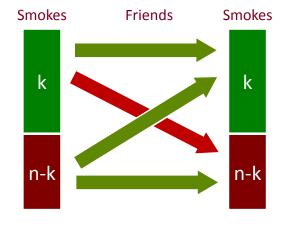
Smokes(Charlie) = 0

Smokes(Dave) = 1

Smokes(Eve) = 0

...

$$\rightarrow 2^{n^2-k(n-k)}$$
 models



• If we know that there are *k* smokers?

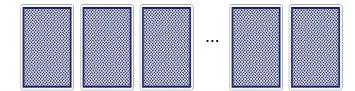
$$\rightarrow \binom{n}{k} 2^{n^2 - k(n-k)}$$
 models

• In total...

$$\rightarrow \sum_{k=0}^{n} \binom{n}{k} 2^{n^2 - k(n-k)}$$
 models

# Playing Cards Revisited

#### Let us automate this:



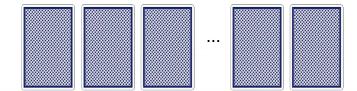
```
\forall p, \exists c, Card(p,c)

\forall c, \exists p, Card(p,c)

\forall p, \forall c, \forall c', Card(p,c) \land Card(p,c') \Rightarrow c = c'
```

# Playing Cards Revisited

#### Let us automate this:



$$\forall p, \exists c, Card(p,c)$$
  
 $\forall c, \exists p, Card(p,c)$   
 $\forall p, \forall c, \forall c', Card(p,c) \land Card(p,c') \Rightarrow c = c'$ 

#SAT = 
$$\sum_{k=0}^{n} {n \choose k} \sum_{l=0}^{n} {n \choose l} (l+1)^k (-1)^{2n-k-l} = n!$$

# Playing Cards Revisited

#### Let us automate this:



$$\forall p, \exists c, Card(p,c)$$
  
 $\forall c, \exists p, Card(p,c)$   
 $\forall p, \forall c, \forall c', Card(p,c) \land Card(p,c') \Rightarrow c = c'$ 

#SAT = 
$$\sum_{k=0}^{n} {n \choose k} \sum_{l=0}^{n} {n \choose l} (l+1)^k (-1)^{2n-k-l} = n!$$

Computed in time polynomial in n

### FO COMPILATION

## **Compilation Rules**

- Lots of preprocessing
- Shannon decomposition/Boole's expansion
- Detect propositional decomposability
- FO Shannon decomposition:

$$\exists \mathbf{X} \subseteq \tau, P(\mathbf{X}) \land \overline{P}(\overline{\mathbf{X}}) \land \beta$$

Simplify  $\beta$  (remove atoms subsumed by P(**X**)) Always deterministic! Ensure automorphic  $\exists$ 

Detect FO decomposability

### FO NNF EXPRESSIVENESS

### Main Positive Result: FO<sup>2</sup>

- $FO^2 = FO$  restricted to two variables
- "The graph has a path of length 10":

```
\exists x \exists y (R(x,y) \land \exists x (R(y,x) \land \exists y (R(x,y) \land ...)))
```

- Theorem: Compilation algorithm to FO ad-DNNF is complete for FO<sup>2</sup>
- Model counting for FO<sup>2</sup> in PTIME domain complexity

# Main Negative Results

#### Domain complexity:

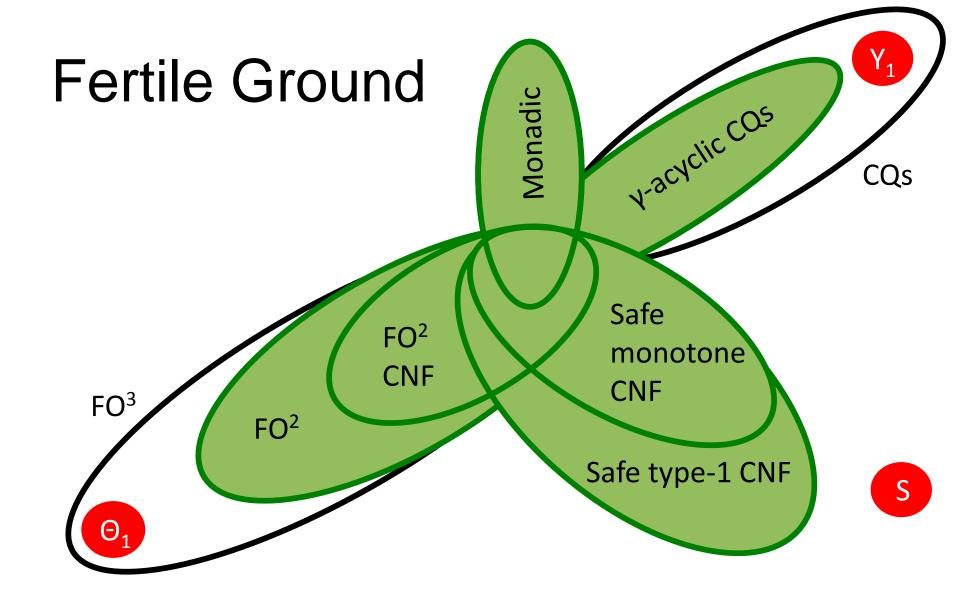
- There exists an FO formula Q s.t. symmetric FOMC(Q, n) is #P<sub>1</sub> hard
- There exists Q in FO<sup>3</sup> s.t. FOMC(Q, n) is #P<sub>1</sub> hard
- There exists a conjunctive query Q s.t. symmetric WFOMC(Q, n) is #P₁ hard
- There exists a positive clause Q w.o. '=' s.t. symmetric WFOMC(Q, n) is #P<sub>1</sub> hard
- Therefore, no FO ad-DNNF can exist 🕾

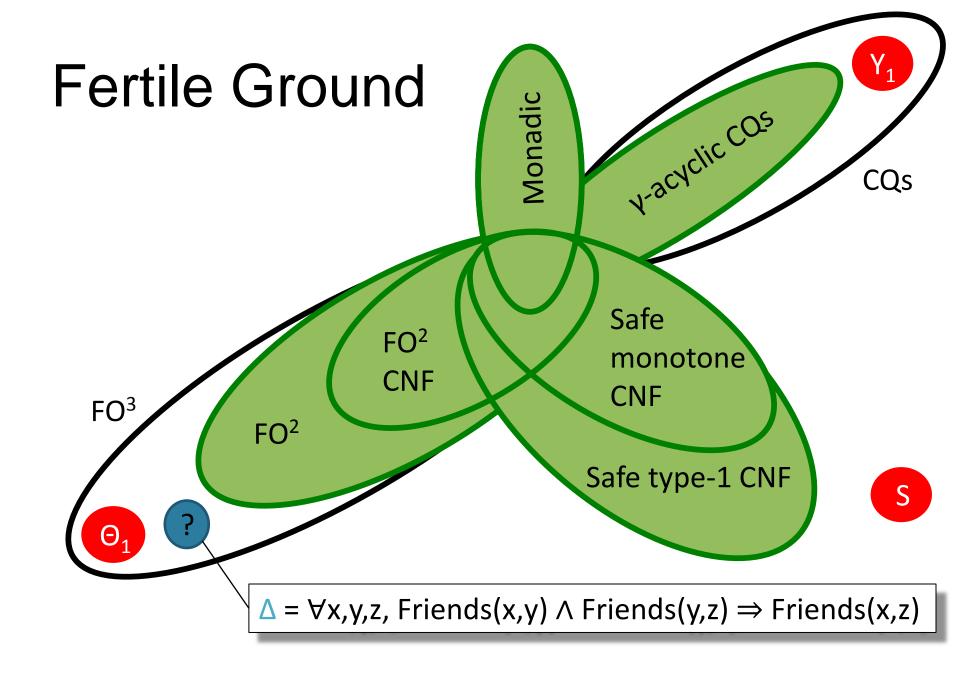
#### **Proof**

**Theorem**. There exists an FO<sup>3</sup> sentence  $\mathbb{Q}$  s.t. FOMC( $\mathbb{Q}$ ,n) is #P<sub>1</sub>-hard

#### **Proof**

- Step 1. Construct a Turing Machine U s.t.
  - U is in #P₁ and runs in linear time in n
  - U computes a #P<sub>1</sub> –hard function
- Step 2. Construct an FO<sup>3</sup> sentence Q s.t. FOMC(Q,n) / n! = U(n)





[VdB; NIPS'11], [VdB et al.; KR'14], [Gribkoff, VdB, Suciu; UAI'15], [Beame, VdB, Gribkoff, Suciu; PODS'15], etc.

#### Other Queries and Transformations

- What if all ground atoms have different weights? Asymmetric WFOMC
- FO d-DNNF complete for all monotone FO CNFs that support efficient CT
- No clausal entailment
- No conditioning

#### Conclusions

- Very powerful already!
- We need to solve this!

#### **THANKS**

#### References

- Cards Example: Guy Van den Broeck. Towards High-Level Probabilistic Reasoning with Lifted Inference, In Proceedings of KRR, 2015.
- First-Order Knowledge Compilation:
   Guy Van den Broeck. Lifted Inference and
   Learning in Statistical Relational Models, PhD
   thesis, KU Leuven, 2013.
- Expressiveness:

   Paul Beame, Guy Van den Broeck, Eric Gribkoff,
   Dan Suciu. Symmetric Weighted First-Order
   Model Counting, In Proceedings of PODS, 2015.