

Markov Logic Networks

Matt Richardson and Pedro Domingos
(2006), Markov Logic Networks,
Machine Learning, 62, 107-136, 2006.

CS 486/686
University of Waterloo
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Outline

- Markov Logic Networks
- Alchemy

Markov Logic Networks

- Bayesian networks and Markov networks:
 - Model uncertainty
 - But propositional representation (e.g., we need one variable per object in the world)
- First-order logic:
 - First-order representation (e.g., quantifiers allow us to reason about several objects simultaneously)
 - But we can't deal with uncertainty
- **Markov logic networks**: combine Markov networks and first-order logic

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Markov Logic

- A logical KB is a set of **hard constraints** on the set of possible worlds
- Let's make them **soft constraints**: when a world violates a formula, it becomes less probable, not impossible
- Give each formula a **weight**: (higher weight \rightarrow stronger constraint)

$$P(\text{world}) \propto e^{\sum \text{weights of formulas it satisfies}}$$

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Markov Logic: Definition

- 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
- Together with a set of constants, it defines a Markov network with
 - One node for each grounding of each predicate in the MLN
 - One feature for each grounding of each formula F in the MLN, with the corresponding weight w

Example: Friends & Smokers

Smoking causes cancer.
Friends have similar smoking habits.

Example: Friends & Smokers

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

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

Example: Friends & Smokers

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

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

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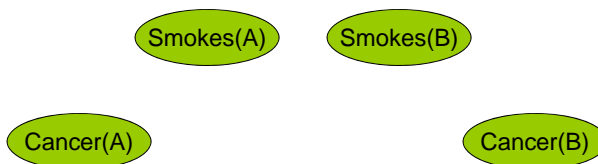
Two constants: **Anna** (A) and **Bob** (B)

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Two constants: **Anna** (A) and **Bob** (B)

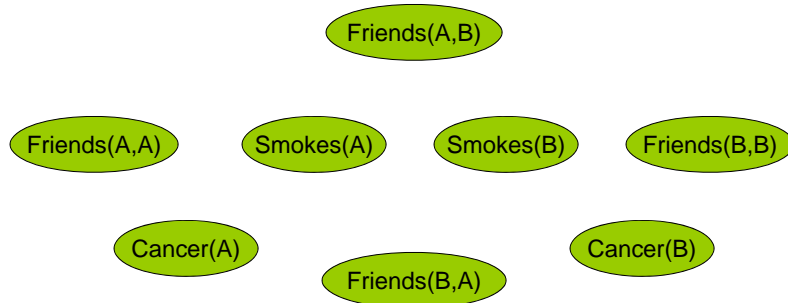


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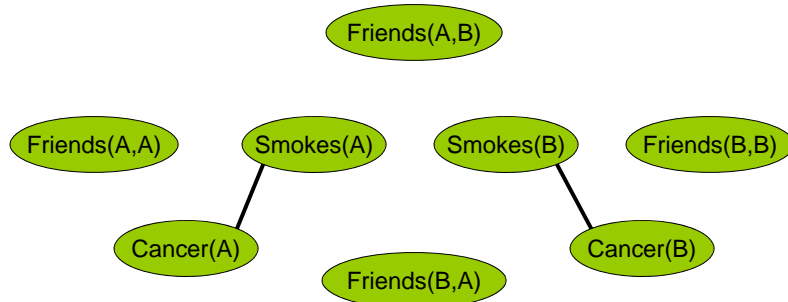
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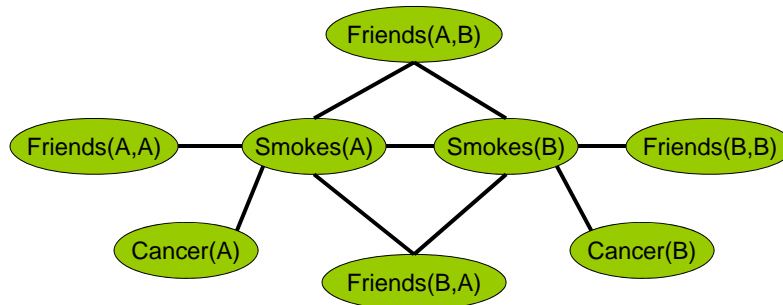
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Markov Logic Networks

- MLN is **template** for ground Markov nets
- Probability of a world x :

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

Weight of formula i

No. of true groundings of formula i in x

- **Typed** variables and constants greatly reduce size of ground Markov net

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Alchemy

- Open Source AI package
- <http://alchemy.cs.washington.edu>
- Implementation of Markov logic networks

- Problem specified in two files:
 - File1.mln (Markov logic network)
 - File2.db (database / data set)

- Learn weights and structure of MLN
- Inference queries

Markov Logic Encoding

- File.mln

- Two parts:
 - Declaration
 - Domain of each variable
 - Predicates
 - Formula
 - Pairs of weights with logical formula

Markov Logic Encoding

- Example declaration
 - Domain of each variable
 - `person = {Anna, Bob}`
 - Predicates:
 - `Friends (person, person)`
 - `Smokes (person)`
 - `Cancer (person)`
 - Example formula
 - `8 Smokes (x) => Cancer (x)`
 - `5 Friends (x, y) => (Smokes (x) <=> Smokes (y))`
- NB: by default, formulas are universally quantified in Alchemy

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Dataset

- File.db
- List of facts (ground atoms)
- Example:
 - `Friends (Anna, Bob)`
 - `Smokes (Anna)`
 - `Cancer (Bob)`

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Syntax

- Logical connective:
 - ! (not), ^ (and), v (or), => (implies), <=> (iff)
- Quantifiers:
 - forall (\forall), exist (\exists)
 - By default unquantified variables are universally quantified in Alchemy
- Operator precedence:
 - ! > ^ > v > => > <=> > forall = exist

Syntax

- Short hand for predicates
 - ! operator: indicates that the preceding variable has exactly one true grounding
 - Ex: `HasPosition(x, y!)`: for each grounding of `x`, exactly one grounding of `y` satisfies `HasPosition`
- Short hand for multiple weights
 - + operator: indicates that a different weight should be learned for each grounding of the following variable
 - Ex: `outcome(throw, +face)`: a different weight is learned for each grounding of `face`

Multinomial Distribution

Example: Throwing dice

Types: `throw = { 1, ... , 20 }`

`face = { 1, ... , 6 }`

Predicate: `Outcome(throw,face)`

Formulas: `Outcome(t,f) ^ f!=f' => !Outcome(t,f')`.

`Exist f Outcome(t,f)`.

Too cumbersome!

Multinomial Distrib.: ! Notation

Example: Throwing dice

Types: `throw = { 1, ... , 20 }`

`face = { 1, ... , 6 }`

Predicate: `Outcome(throw,face!)`

Formulas:

Semantics: Arguments without "!" determine args with "!".

Only one face possible for each throw.

Multinomial Distrib.: + Notation

Example: Throwing biased dice

Types: `throw = { 1, ... , 20 }`

`face = { 1, ... , 6 }`

Predicate: `Outcome(throw,face!)`

Formulas: `Outcome(t,+f)`

Semantics: Learn weight for each grounding of args with "+".

Text Classification

`page = { 1, ... , n }`

`word = { ... }`

`topic = { ... }`

`Topic(page,topic!)`

`HasWord(page,word)`

`Links(page,page)`

`HasWord(p,+w) => Topic(p,+t)`

`Topic(p,t) ^ Links(p,p') => Topic(p',t)`

Next Class

- Applications of Markov Logic Networks