

Expressivity Analysis for PL-Languages

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Semantics-based Expressivity Analysis

The Problem

“Alphabet soup” (L.Getoor): Prism, SLP, RBN, PRM, BLP, MLN, Blog, ...

Questions:

- Where are these languages similar?
- Where are these languages different?
- What are the particular strengths/weaknesses of language XYZ?

First issue to investigate:

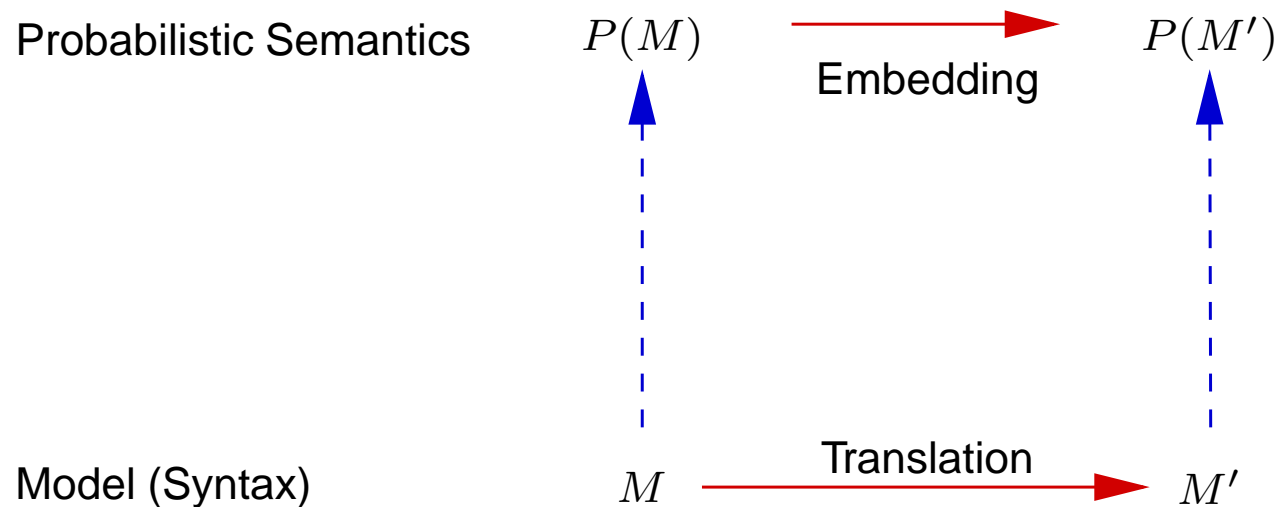
- What is the expressive power of the different languages?

Later:

- What is the complexity of inference?
- What is the complexity of learning?

Elements of a Solution

- Goal: establish general framework with re-usable components for expressivity analysis
- Find common semantic ground
- Consider *translations* of (syntactic) models and *embeddings* of their semantics.
- A language L' is *at least as expressive as* a language L , if each L -model M can be translated into an L' -model M' , so that the semantics of M' “contains” the semantics of M .



Common Semantic Ground: Multi-valued Herbrand Interpretations

PL-languages define distributions for random variables that can be written as ground atoms:

blood_pressure(tom) *sister(susan,tom)* *genotype(mother(paul))*
blood_pressure(susan) *sister(susan,paul)* *genotype(father(paul))*
...

With each relation symbol is associated a (finite) state space:

states(blood_pressure)={ high, normal, low}
states(sister)={ true, false}
states(genotype)={ AA, Aa, aa}

Herbrand Interpretation: assignment of a truth value to all ground atoms constructible from a vocabulary S of relation, function, and constant symbols.

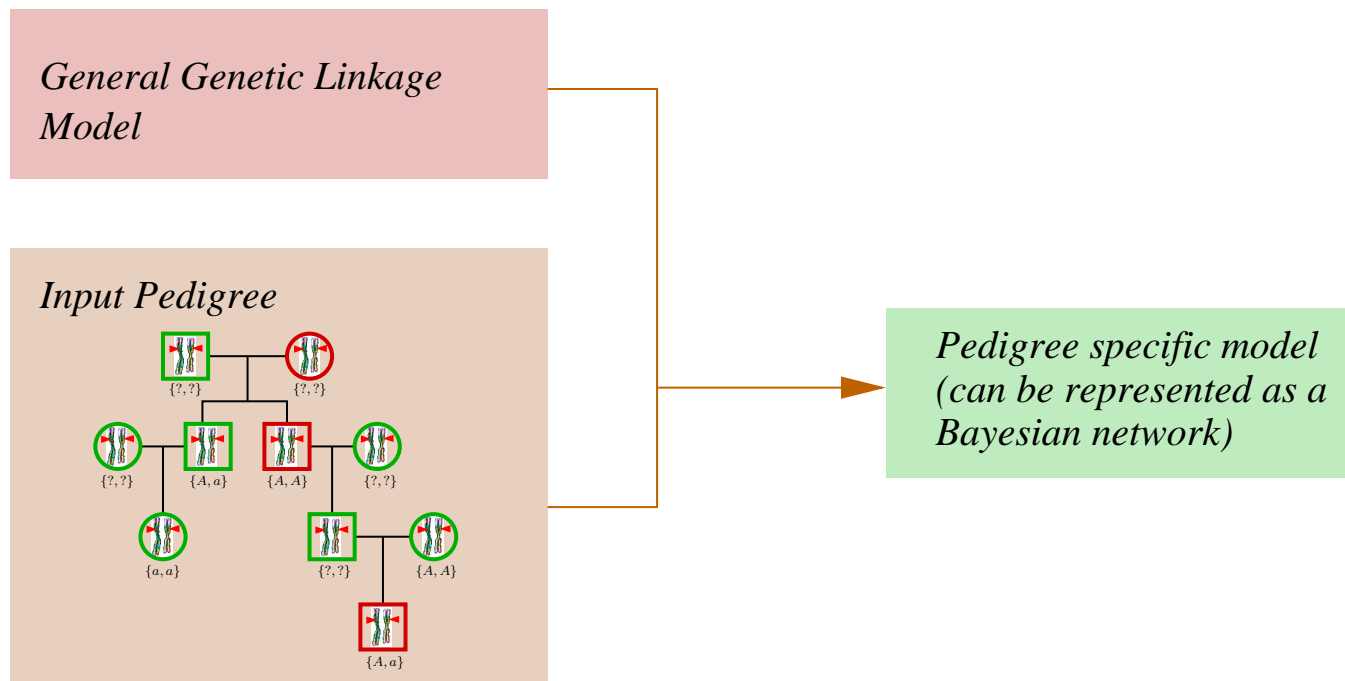
Multi-valued Herbrand Interpretation: assignment of a state to all ground atoms constructible from a vocabulary S of relation, function, and constant symbols.

PL-model: defines a probability distribution over all Multi-valued Herbrand Interpretations for a given vocabulary S .

Any PL-model can be represented by an ordinary Bayesian network. Are PL-languages just shorthand notations for large Bayesian networks?

Modularity of Representations

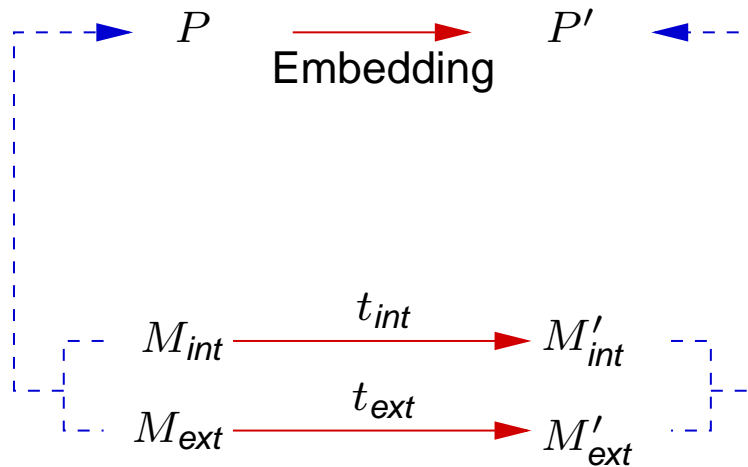
The power and usefulness of PL-languages derives from the fact that they split the specification of a complex model into a generic (*intensional*) and a domain-specific (*extensional*) part:



A (preliminary) analysis of several languages:

	Intensional	Extensional
RBN	rbn	Input Structure
PRM	prm	Skeleton Structure
BLP	intensional part	extensional part
MLN	mln	constants
Prism	ground atoms with without msw's in SLD tree	

Updated plan:



Formalization

Embeddings

P : probability distributions over $MVHI(S)$

P' : probability distributions over $MVHI(S')$

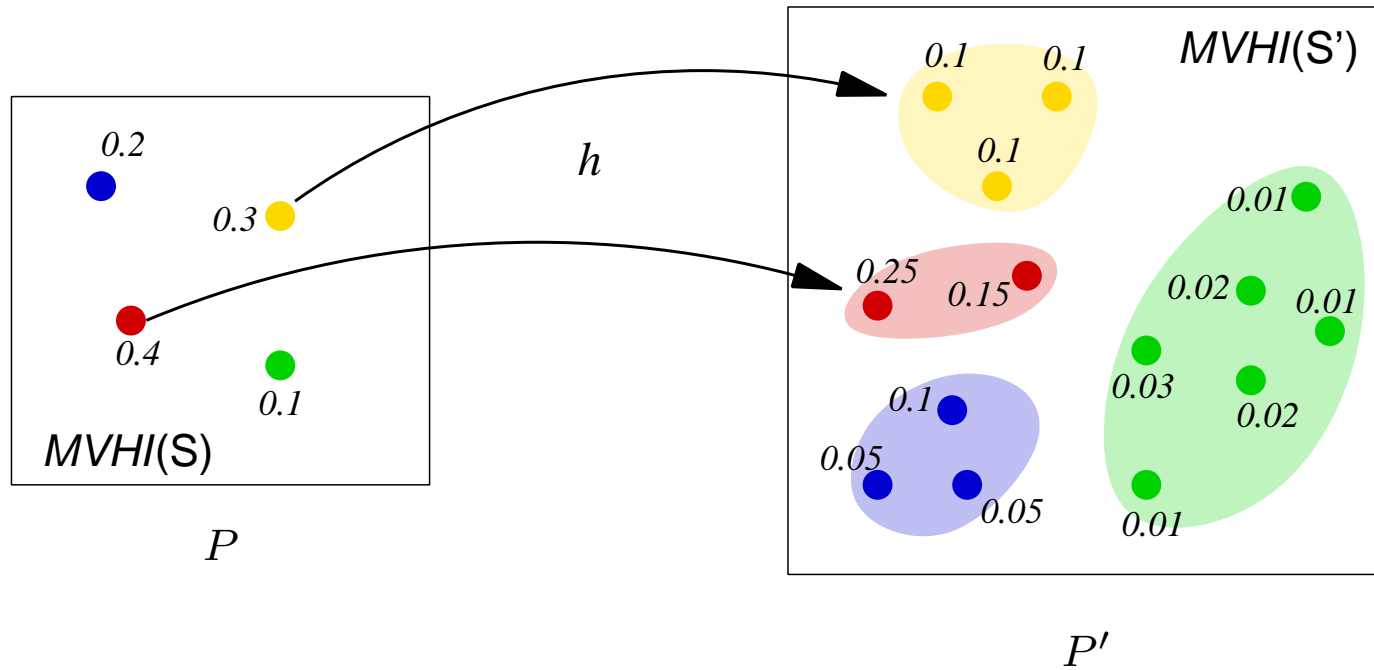
An embedding of P in P' is a mapping

$$h : MVHI(S) \mapsto 2^{MVHI(S')}$$

such that for all $w, w' \in MVHI(S)$:

$$P(w) = P'(h(w)) \text{ and } h(w) \cap h(w') = \emptyset$$

Write $P \preceq P'$ if there is such an embedding.



If $P \preceq P'$, then every probabilistic query about P can be answered from the model P' (one can consider weaker forms of embeddings, so that only restricted types of queries for P are supported by P').

Putting Everything Together...

Language L' is at least as expressive as L , $L \preceq L'$, if

$$\exists t_{int} \forall M_{int} \exists t_{ext} \forall M_{ext} \quad P(M_{int}, M_{ext}) \preceq P(t_{int}(M_{int}), t_{ext}(M_{ext}))$$

Example Result

$$MLN \preceq RBN \quad (\text{precisely: } MLN \preceq_c RBN)$$