

# Markov Logic Networks for Natural Language Question Answering

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# Task: Answering Elementary Science Questions (i.e. Non-Factoid QA)

Example Question:

Question: Living things depend on energy from

(A) the Sun (B) the Moon (C) soil (D) water

As T/F: Living things depend on energy from the Sun.

# Knowledge Base:

Sentences parsed with AI2's FOL parser from Clark (2014) into IF-THEN rules:

Example:

Growing thicker fur in the winter helps some animals to stay warm

$$\begin{aligned} & isa(g, \text{grow}), isa(a, \text{some\_animals}), isa(f, \text{thicker\_fur}), \\ & isa(w, \text{the\_winter}), agent(g, a), object(g, f), in(g, w) \\ & \Rightarrow \exists s, r : isa(s, \text{stays}), isa(r, \text{warm}), \\ & \quad enables(g, s), agent(s, a), object(s, r) \end{aligned} \tag{1}$$

# Questions Parsed:

Questions are parsed similarly:

Question (as T/F):

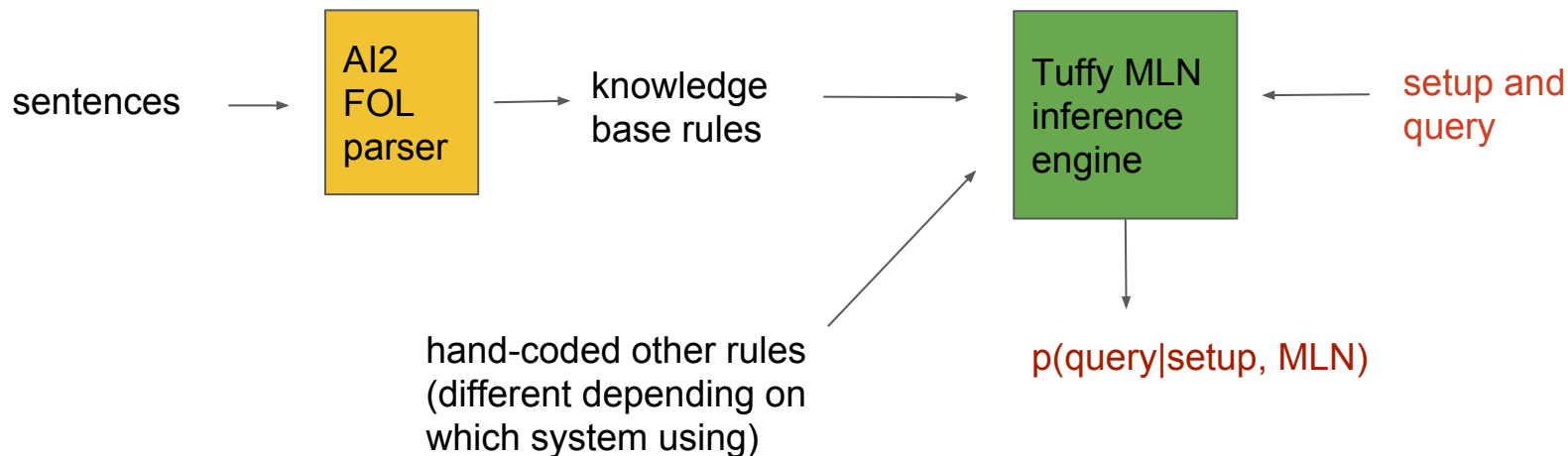
A fox grows thick fur as the season changes. This helps the fox to keep warm

$setup : isa(F, fox), isa(G, grows), isa(T, thick\_fur),$

$agent(G, F), object(G, T)$

$query : isa(K, keep\_warm), enables(G, K), agent(K, F)$

# Main Architecture (as I understand it...)



# 1st Set-up: FOL-MLN

$$\begin{aligned} & isa(g, grow), isa(a, some\_animals), isa(f, thicker\_fur), \\ & isa(w, the\_winter), agent(g, a), object(g, f), in(g, w) \\ & \Rightarrow \exists s, r : isa(s, stays), isa(r, warm), \\ & \quad enables(g, s), agent(s, a), object(s, r) \end{aligned} \quad (1)$$

Rules in:

- KB rules
- semantic rules ( $cause(x,y) \rightarrow effect(y,x)$ , etc)
- lexical alignment “rules” (from external algorithm)

$$query : isa(K, keep\_warm), enables(G, K), agent(K, F)$$

GOAL: train MLN and find probability of answer choice, (a separate grounding??): i.e.

$$result(query) = true$$

**Big Issue:** Too many groundings

## 2nd Set-up: ER-MLN

$isa(G, grow), isa(A, some\_animals), isa(F, thicker\_fur),$   
 $isa(W, the\_winter), agent(G, A), object(G, F), in(G, W)$   
 $\Rightarrow isa(S, stays), isa(R, warm),$   
 $enables(G, S), agent(S, A), object(S, R)$

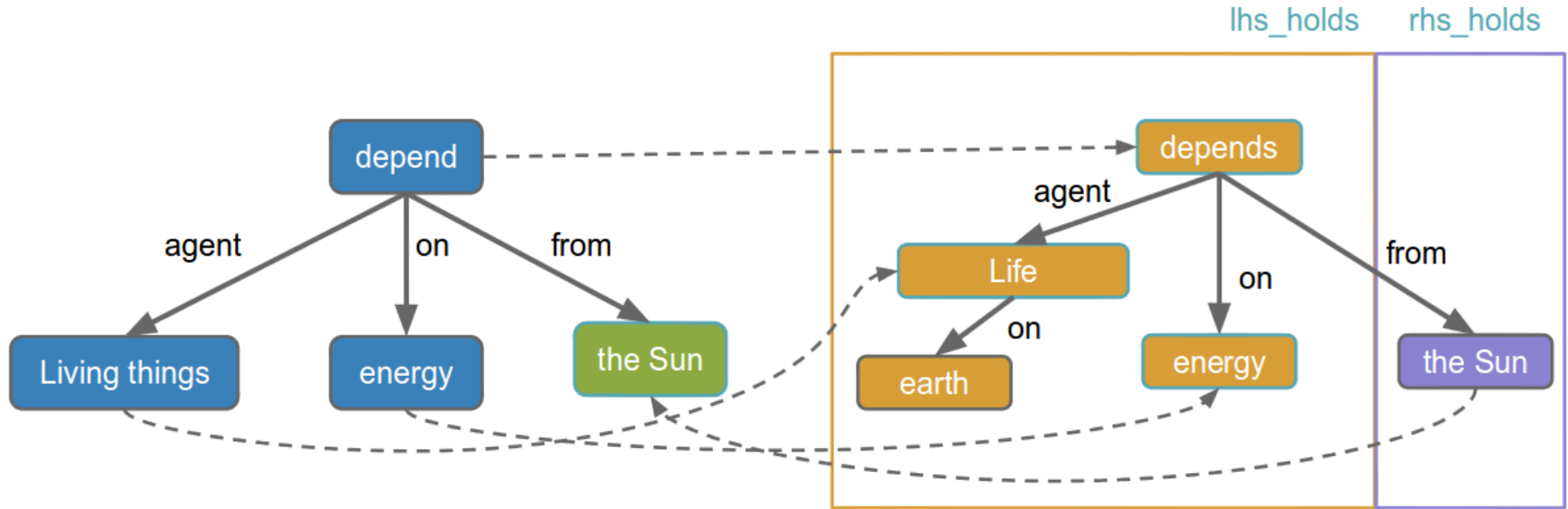
Rules in:

- KB rules - **NOW with no variables**
- semantic rules ( $cause(x,y) \rightarrow effect(y,x)$ , etc)
- lexical alignment “rules” (from external algorithm)
- **NEW -- Equivalence/Resolution rules:**  $sameAs()$  predicate which interacts with others with hand-coded rules (p.4)
- **NEW -- Partial Match rules:**  $J, K, L, \dots \rightarrow R$  you also have  $J \rightarrow R, K \rightarrow R, L \rightarrow R, \dots$

**GOAL:** train MLN and find probability of answer choice, (a separate grounding??): i.e.  $result(query) = true$

**Big Issue:** Brittle to syntactic differences (but better than FOL-MLN) and problems with words like “bat” (i.e. baseball vs. animal)

New idea - use MLN to do Alignment of graphical structure:





## 3rd Set-up: PRALINE

$holds(Grow), holds(Animals), holds(Fur),$   
 $holds(Winter) \Rightarrow holds(Stays), holds(Warm)$

“Rules” in: (very different, I think...)

- **NEW:** want - Nodes (words?) and edges (alignments between nodes) of  $G$ , frame as rules:
  - $Node(nodeid), edge(nodeid, nodeid, label), setup(nodeid), query(nodeid), inLHS(nodeid), inRHS(nodeid)...$
- **NEW: Graph alignment rules** -- codes up the lexical alignment... ~ transitivity, etc.
- **NEW: Inference rules:** -- to get new rules...  $holds(x), aligns(x, y) \Rightarrow holds(y)$
- (??old rules??)

GOAL: ??? like:  $P(holds(Stays), holds(Warm) \mid holds(Fox), holds(Grows), holds(Fur), M)$

Success!