

Formalizing AMR Inference via Hybrid Logic Tableaux

CL Masters Thesis Defense

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Introduction

- ▶ Semantic representation:
 - ▶ Capture meaning of natural language content.
 - ▶ Designed for manipulation via software.
- ▶ Abstract Meaning Representation (AMR):
 - ▶ Graph-based (DAG), nodes are *concepts*, edges are *relations*.
 - ▶ Built on predicative core of a sentence.
 - ▶ Ignores syntactic differences between equivalent sentences.
 - ▶ PropBank framesets are used for concepts (entities, events, properties, states).

Introduction

(Basic) AMR is reductionistic. This is awesome for:

- ▶ Annotation (esp. by non-experts).
- ▶ Semantic parsing (smaller target space).

This is not awesome for:

- ▶ Representing and recovering fine-grained meaning.
- ▶ Automating reasoning/inference.

Introduction

- ▶ The trade-off between ease of generation/use and rich expressivity/inferentiability is at least as old as computing.
- ▶ AMR has made a choice that works well in data-driven NLP.
- ▶ However AMR can bridge this gap:
- ▶ AMR already does this in a modular way with extensions.
- ▶ Some of these extensions give afford interpretation in first-order logic:
 - ▶ Automated inference for logics is a rich area with lots of tools.
 - ▶ This is where we come in.

Motivation

“Why do we need formal methods? Can’t state-of-the-art language models do this already?”

Short answer: Not really, and even if they could:

- ▶ Statistically driven techniques are unnecessarily expensive for formal inference.
- ▶ Increasing need for ability to guarantee/verify properties of software:
 - ▶ Does the software give us the right *type* of result for an input?
 - ▶ Bias in NLP.
- ▶ Machine learning (by itself) does not lend itself well to this.

Approach

- ▶ Combine two AMR extensions for richer interpretation:
 - ▶ Scope and quantification (?)
 - ▶ Tense and aspect (?)
- ▶ Interpret these extended AMR into a logic that handles quantification and tense.
- ▶ Develop tableau methods for this logic:
 - ▶ General method for proving/disproving sentences in the logic.
 - ▶ Restricted method for checking if sentence holds in some model.

AMR with Scope and Quantification

- ▶ Disambiguates scope.
- ▶ Annotates central predicate and its arguments.
- ▶ Clearest path for AMR \rightarrow standard first-order predicate logic.

AMR with Tense and Aspect

- ▶ Standard AMR structure.
- ▶ Central predicate annotated for:
 - ▶ Aspect.
 - ▶ Event time.
 - ▶ Reference time.

Combined Extensions

- ▶ Assume each AMR has information from both extensions.
- ▶ Attach tense and aspect information to central predicate node.
- ▶ Extract a tense-sensitive FOPL representation (details later).

Modal Logic

- ▶ Propositional logic lets us form statements like $p \wedge (q \vee \neg r)$.
- ▶ Modal propositional logic extends propositional logic with an operator \Diamond , read as “possible”. i.e. it is not possible that p and $\neg p$ are the case would be:

$$\neg \Diamond(p \wedge \neg p)$$

- ▶ (More) formal meaning of \Diamond : There is a *possible world* where p is true, and this possible world is *accessible* from the current one.
- ▶ The problem: the “current world” is an implicit notion dependent on context. Is there something more expressive?

Hybrid Logic

- ▶ Idea: take propositional modal logic, and add an operator @, that lets us know which world we're referring to.
- ▶ p or r is possible at world i : $@_i \Diamond(p \vee r)$
- ▶ In the above proposition i is called a nominal since it *names* some/is true at exactly one world.
- ▶ Everything true at the nominal j is true at the nominal i (they name the same world): $@_i j$

Hybrid Logic Variants

- ▶ Hybrid tense logic:
 - ▶ Two tense modalities: $\langle F \rangle$ and $\langle P \rangle$
 - ▶ World j is in the past of world i : $@_i Pj$
- ▶ Quantified hybrid logic:
 - ▶ Hybrid logic with first-order quantifiers, relation, and function symbols.
 - ▶ At n (now) there is a person for whom it is possible to own a car:

$$@_n(\exists x)(Person(x) \wedge (\exists y)(Car(y) \wedge \Diamond Afford(x, y)))$$

Hybrid Logic Variants

A problem

$$@_n(\exists x)(Person(x) \wedge (\exists y)(Car(y) \wedge \Diamond Afford(x, y)))$$

What's the domain of quantification?

- ▶ Option 1: The domain is the same at every world: *There is someone (out of all people all people at all worlds) who can afford some car (out of all cars at all worlds).*
- ▶ Option 2: Each world has a different domain: *There is someone (out of everyone in this world) who can afford some car (out of all cars in this world).*

The first option in quantified hybrid (or modal) logic is called possibilist quantification, the second is called actualist. When worlds are interpreted as times, the former is called eternalist quantification, and the latter is called presentist. We will use presentist quantification.

First-Order Hybrid Tense Logic

First-order Hybrid Tense Logic (*FHTL*) is:

- ▶ Quantified hybrid logic, with tense modalities F and P instead of \Diamond .
- ▶ Presentist quantification:
 - ▶ Each world has its own domain.
 - ▶ Quantification is domain sensitive.

First-Order Hybrid Tense Logic

In logic models are interpretations of symbols and constants in the language. An *FHTL* model \mathfrak{M} is a tuple

$$(T, \mathcal{R}, (D_t)_{t \in T}, I_{nom}, (I_t)_{t \in T})$$

Where:

- ▶ T is a set of times/worlds.
- ▶ \mathcal{R} is the binary accesibility relation over times.
- ▶ D_t is the domain of a time t
- ▶ I_{nom} assigns nominals to worlds.
- ▶ I_t interprets the value of terms at a time t .

The satisfiability of a formula with free variables depends on a variable assignment function. A formula that does not depend on variable assignment (every variable is bound by a quantifier) is called a *sentence*.

First-Order Hybrid Tense Logic

- ▶ Two main tasks for a sentence s :
 - ▶ Theorem proving – is s true regardless of the model?
 - ▶ Model checking – is s true in some model \mathfrak{M} ?
- ▶ For both we use versions of the tableau method.

FHTL Tableau Example

- (1)
- (2) $@_s(\exists x)[P((\exists y)[f(x, y) = f(y, x)]) \vee \neg(\exists z)[x = z]]$
- (3) $@_sP((\exists y)[f(s_1, y) = f(y, s_1)]) \vee \neg(\exists z)[s_1 = z]$
- (4) $@_sP((\exists y)[f(s_1, y) = f(y, s_1)])$ $@_s\neg(\exists z)[s_1 = z]$
- (5) $@_sPt$ $@_s\neg[s_1 = s_1]$
- (6) $@_t(\exists y)[f(s_1, y) = f(y, s_1)]$ $@_s[s_1 = s_1]$
- (7) \dots \otimes

Model Checking Example

- *Every computer will be located at a desk.*

- AMR with quantification and tense:

```
(s / scope
  :pred (b / be-located-at-91 :ongoing -
                                :complete +
                                :time (a / after
                                         :op1 (n / now))
  :ARG0 (c / computer)
  :ARG1 (d / desk
         :quant (e / every)))
:ARG0 d
:ARG1 c)
```

- *FHTL* translation:

$$@_{now}(\forall y)[desk(y) \rightarrow (\exists x)[computer(x) \wedge F(be-located-at-91(x, y))]]$$

Model Checking Example

Define a small *FHTL* model $\mathfrak{M} = (T, \mathcal{R}, (D_t)_{t \in T}, I_{nom}, (I_t)_{t \in T})$ where:

$$T = \{\text{yesterday}, \text{now}, \text{tomorrow}\}$$

$$\mathcal{R} = \{(\text{yesterday}, \text{now}), (\text{now}, \text{tomorrow}), (\text{yesterday}, \text{tomorrow})\}$$

$$I_{nom} = \{(y, \text{yesterday}), (n, \text{now}), (t, \text{tomorrow})\}$$

$$D_{\text{yesterday}} = \{\text{computer}_1, \text{desk}_1\}$$

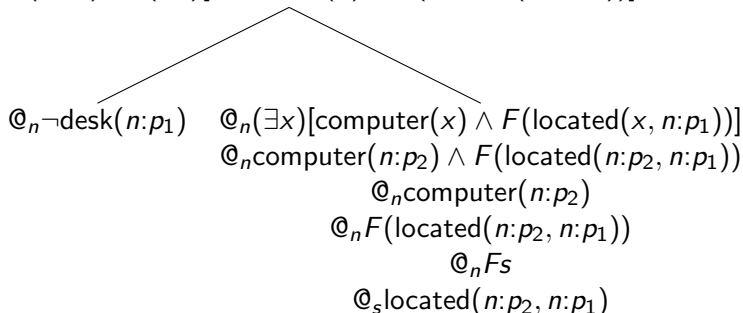
$$D_{\text{now}} = \{\text{computer}_1, \text{computer}_2, \text{desk}_1, \text{desk}_2, \text{desk}_3\}$$

$$D_{\text{tomorrow}} = \{\text{computer}_1, \text{computer}_2, \text{desk}_1, \text{desk}_2\}$$

Model Checking Example

$$@_n(\forall y)[\text{desk}(y) \rightarrow (\exists x)[\text{computer}(x) \wedge F(\text{located}(x, y))]]$$

$$@_n\text{desk}(n:p_1) \rightarrow (\exists x)[\text{computer}(x) \wedge F(\text{located}(x, n:p_1))]$$



Where we assign $s = t = \text{tomorrow}$, $n:p_2 = \text{computer}_1$ and $n:p_1 = \text{desk}_2$, or $n:p_2 = \text{computer}_2$ and $n:p_1 = \text{desk}_1$ we see that \mathfrak{M} satisfies the *FHTL* sentence.

Extraction

Interpretation

References I