

— *Master's Thesis Presentation* —

Prototyping NLU Pipelines

A Type-Theoretical Framework

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presented virtually due to COVID-19

December 16, 2020

Disclaimer

Some of slides have been adapted from previous presentations by me, often about joint research with other people.

Natural Language Understanding (NLU)

My definition:

NLU means translating natural language into a formal semantic representation.

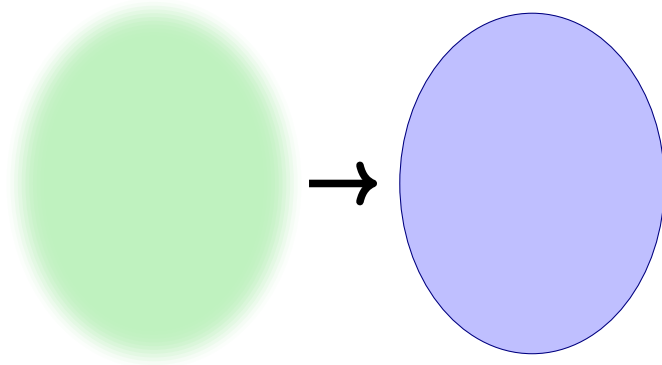
“How many people live in Slovakia?” \rightsquigarrow *“5.458 million”*

“Do more people live in Slovakia than in Thailand?” \rightsquigarrow ???

Method of Fragments

Natural Language

Logic



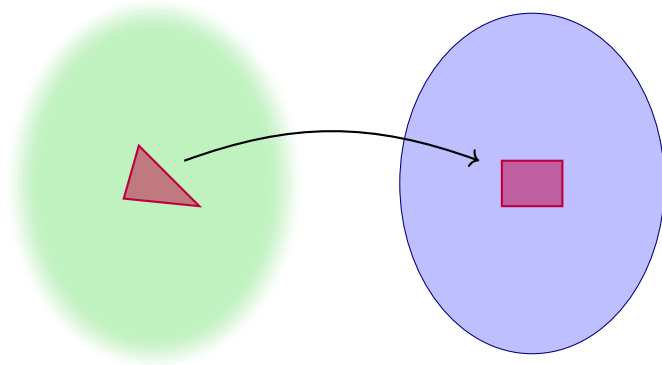
How do we get from messy language to formal logic?

Montague [Mon70]: Look at a “nice” subset and map into logic.

Method of Fragments

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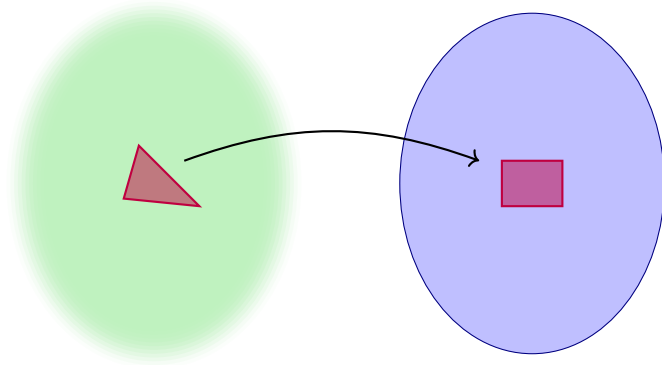
How do we get from messy language to formal logic?

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"Ahmed paints and Berta is quiet."

"Ahmed doesn't paint."

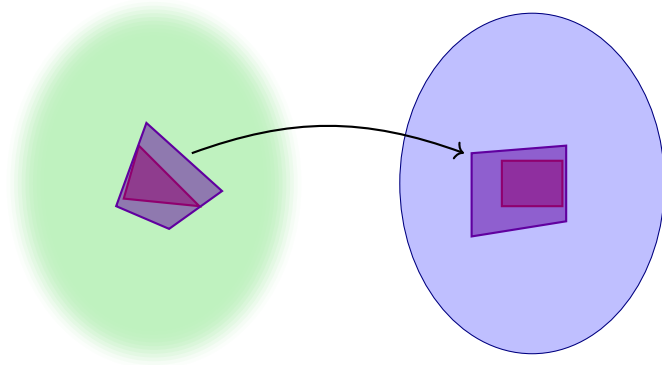
$p(a) \wedge q(b)$

$\neg p(a)$

Method of Fragments

Natural Language

Logic



“Every student paints and is quiet.”

“Nobody paints.”

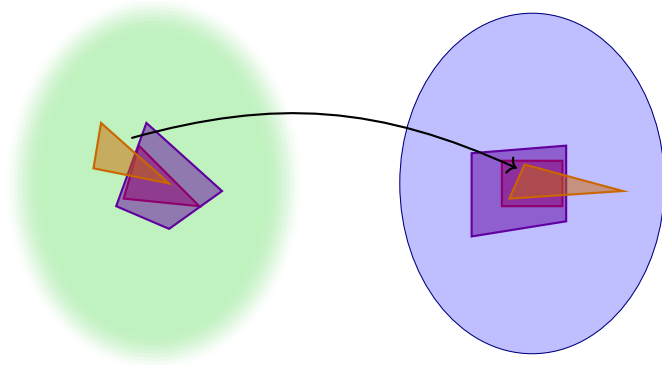
$\forall x.s(x) \Rightarrow (p(x) \wedge q(x))$

$\neg \exists x.p(x)$

Method of Fragments

Natural Language

Logic



“Ahmed isn’t allowed to paint.”

“Ahmed and Berta must paint.”

$\neg \Diamond p(a)$

$(\Box p(a)) \wedge \Box p(b)$

Method of Fragments

Hand-waving is problematic:

“Ahmed paints. He is quiet.” $\rightsquigarrow^?$ $p(a) \wedge q(a)$

Montague: Specify

- grammar,
- target logic,
- semantics construction.

fixes NL subset

maps parse trees to logic

Example from [Mon74]

- T11. If $\phi, \psi \in P_t$ and ϕ, ψ translate into ϕ', ψ' respectively, then ϕ **and** ψ translates into $[\phi \wedge \psi]$, ϕ **or** ψ translates into $[\phi \vee \psi]$.
- T12. If $\gamma, \delta \in P_{IV}$ and γ, δ translate into γ', δ' respectively, then γ **and** δ translates into $\hat{x}[\gamma'(x) \wedge \delta'(x)]$, γ **or** δ translates into $\hat{x}[\gamma'(x) \vee \delta'(x)]$.
- T13. If $\alpha, \beta \in P_T$ and α, β translate into α', β' respectively, then α **or** β translates into $\hat{P}[\alpha'(P) \vee \beta'(P)]$.

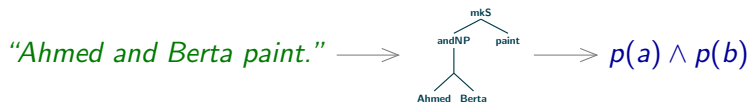
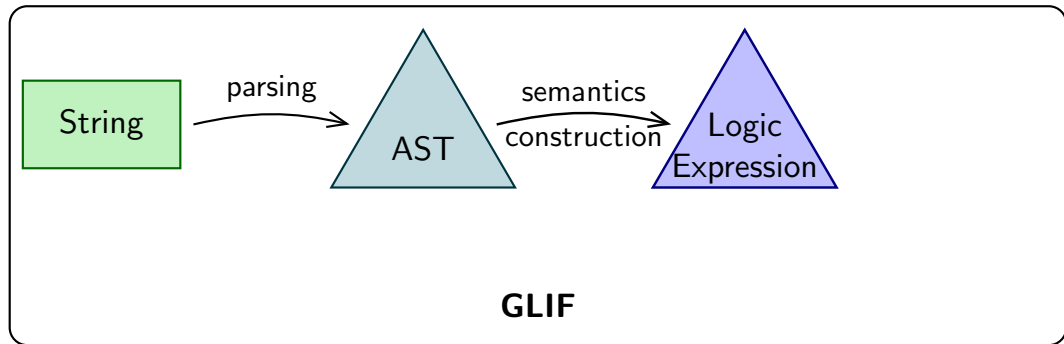
Claim: That doesn't scale well \rightsquigarrow **We need prototyping!**

NLU Prototyping

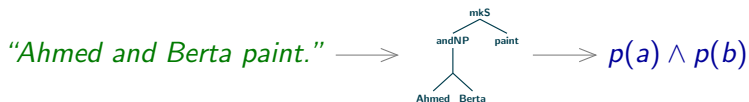
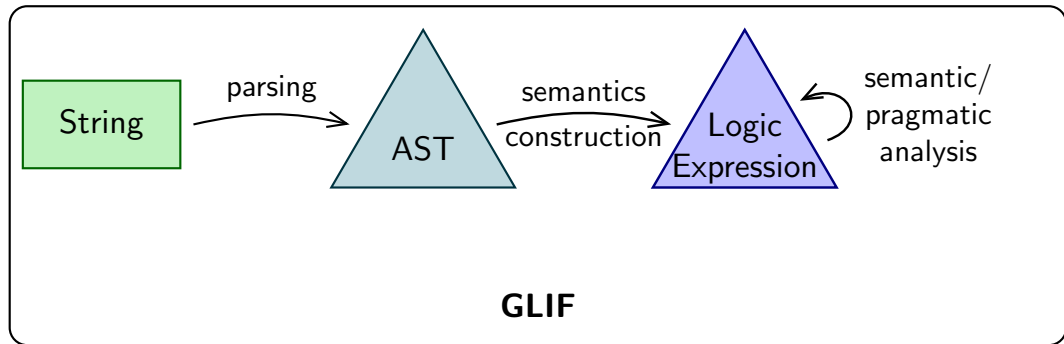
- Traditionally done in Prolog/Haskell
 - requires a lot of work
- A dedicated framework might be better
 - only partial solutions exist
- Can we combine existing partial solutions?
 - ↪ GLIF

Research Question

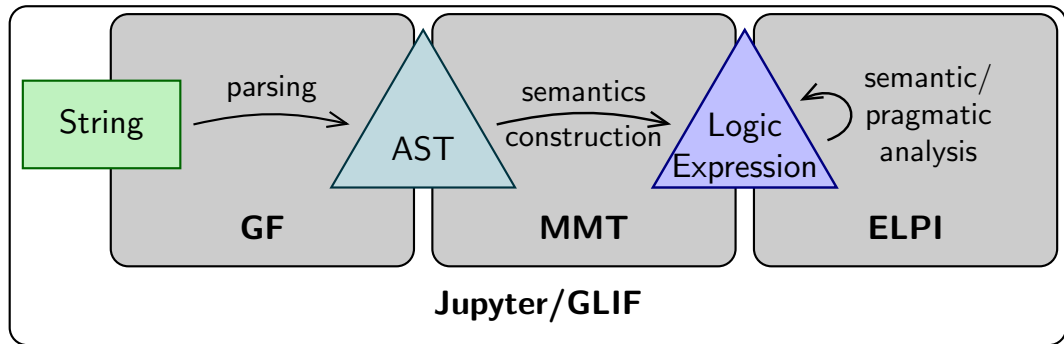
GLIF: Grammatical Logical Inference Framework



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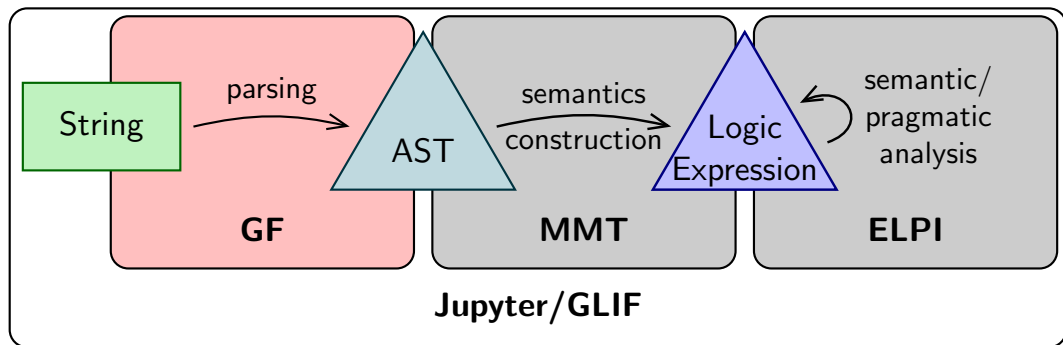
GF (= **grammar** framework)

+ **MMT** (= **logic** framework)

+ **ELPI** (= **inference** framework)

= **GLIF** (= **natural language understanding** framework)

Components of GLIF: GF



Components of GLIF: Grammatical Framework [GF]

- Specialized for developing natural language grammars

- Separates abstract and concrete syntax

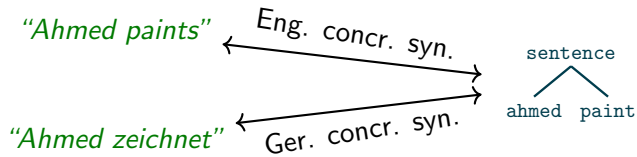
sentence : NP -> VP -> S; *--abstract*

sentence np vp = np.s ++ vp.s!np.n; *--concrete*

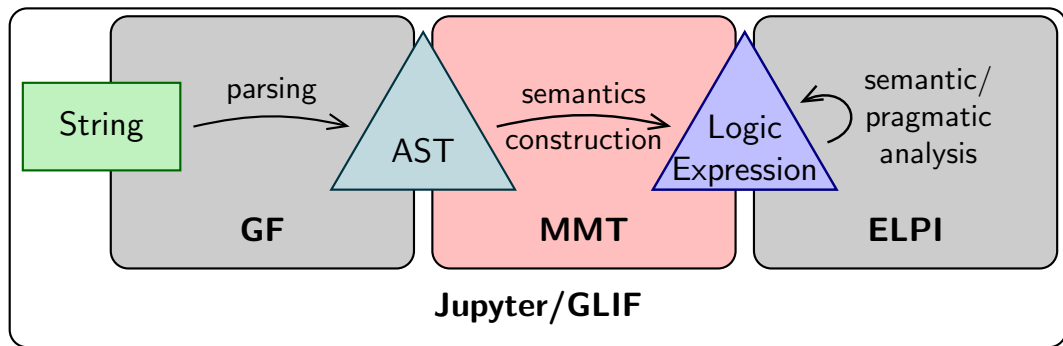
- Abstract syntax based on LF

- Comes with large library

≥ 36 languages



Components of GLIF: MMT



Components of GLIF: MMT

- Modular logic development and knowledge representation
- Not specialized in one logical framework
- We will use MMT to:
 - represent abstract syntax
 - specify target logic and discourse domain theory
 - specify semantics construction

we use LF

Components of GLIF: MMT

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 - **represent abstract syntax**
 - specify target logic and discourse domain theory
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we use LF

GF

```
cat
  NP; VP; S;
fun
  sentence :
    NP -> VP -> S;
```



MMT

```
NP : type
VP : type
S  : type
sentence :
  NP → VP → S
```

Components of GLIF: MMT

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 - represent abstract syntax
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we use LF

Logic Syntax

```
o : type //propositions
¬ : o → o
∧ : o → o → o
∨ : o → o → o

ι : type //individuals
∀ : (ι → o) → o
∃ : (ι → o) → o
```

Discourse Domain Theory

```
paint : ι → o
quiet : ι → o
ahmed : ι
berta : ι
```

idea: $\forall f$ or $\forall \lambda x.f(x)$
instead of $\forall x.f(x)$

Components of GLIF: MMT

- Modular logic development and knowledge representation
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Semantics Construction

map symbols in abstract syntax to terms in logic/domain theory

Simple setting

S	\mapsto o
NP	\mapsto ι
VP	\mapsto ι \rightarrow o
sentence	\mapsto $\lambda n. \lambda v. v$ n
ahmed	\mapsto ahmed

More advanced

NP	\mapsto (ι \rightarrow o) \rightarrow o
sentence	\mapsto $\lambda n. \lambda v. n$ v
everyone	\mapsto $\lambda p. \forall \lambda x. p$ x
berta	\mapsto $\lambda p. p$ berta

Example: Parsing + Semantics Construction

“Ahmed and Berta paint”

↓_{parsing}

sentence (andNP ahmed berta) paint

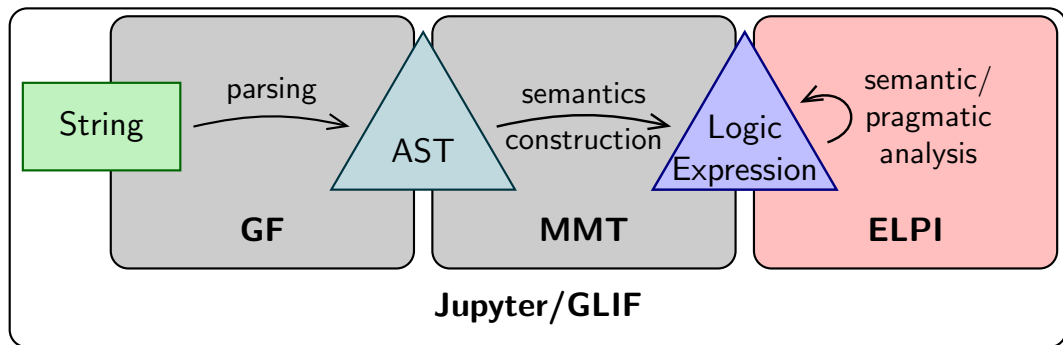
↓_{semantics construction}

$(\lambda n. \lambda v. n \ v) \ ((\lambda a. \lambda b. \lambda p. a \ p \wedge b \ p) \ (\lambda p. p \ \text{ahmed}) \ (\lambda p. p \ \text{berta})) \ \text{paint}$

↓ _{β -reduction}

$\text{paint} \ \text{ahmed} \wedge \text{paint} \ \text{berta}$

Components of GLIF: ELPI



Components of GLIF: ELPI

- Implementation and extension of λ Prolog
- MMT can generate logic signatures
- Generic inference/reasoning step after semantics construction
- Goal: Use it for semantic/pragmatic analysis

\approx *Prolog + HOAS*

MMT

```
o : type //propositions
¬ : o → o
∧ : o → o → o
∨ : o → o → o

ι : type //individuals
∀ : (ι → o) → o
∃ : (ι → o) → o
```

ELPI

```
kind o type.
not : o -> o.
and : o -> o -> o.
or  : o -> o -> o.

kind i type.
type forall (i -> o) -> o.
type exists (i -> o) -> o.
```

Example: Epistemic Q&A

John knows that Mary or Eve knows that Ping has a dog. (S_1)

Mary doesn't know if Ping has a dog. (S_2)

Does Eve know if Ping has a dog? (Q)

$$S_1 = \Box_{\text{john}}(\Box_{\text{mary}}hd(\text{ping})) \vee \Box_{\text{eve}}hd(\text{ping})$$

$$S_2 = \neg((\Box_{\text{mary}}hd(\text{ping})) \vee \Box_{\text{mary}}\neg hd(\text{ping}))$$

$$Q = (\Box_{\text{eve}}hd(\text{ping})) \vee \Box_{\text{eve}}\neg hd(\text{ping})$$

$$S_1, S_2 \vdash_{S5_n} Q \quad \rightsquigarrow \quad \text{yes}$$

$$S_1, S_2 \vdash_{S5_n} \neg Q \quad \rightsquigarrow \quad \text{no}$$

$$\text{else} \quad \rightsquigarrow \quad \text{maybe}$$

“The trophy doesn’t fit in the brown suitcase because it’s too big.” [LDM12]

Semantic/Pragmatic Analysis

“The trophy doesn’t fit in the brown suitcase because it’s too big.” [LDM12]
“The trophy doesn’t fit in the brown suitcase because it’s too small.” [LDM12]

Semantic/Pragmatic Analysis

“The trophy doesn’t fit in the brown suitcase because it’s too big.” [LDM12]
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“The ball has a radius of 2m.”
“The ball has a mass of 2m.”

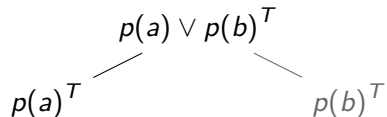
“We saw her duck.”

↪ semantics construction creates preliminary semantic representation(s) that gets refined by the semantic/pragmatic analysis

Example: Tableaux Machine [KK03]

- Can use tableaux for model generation
- Tableau machine: pick “best” branch as model and continue there with next sentence
like a human?

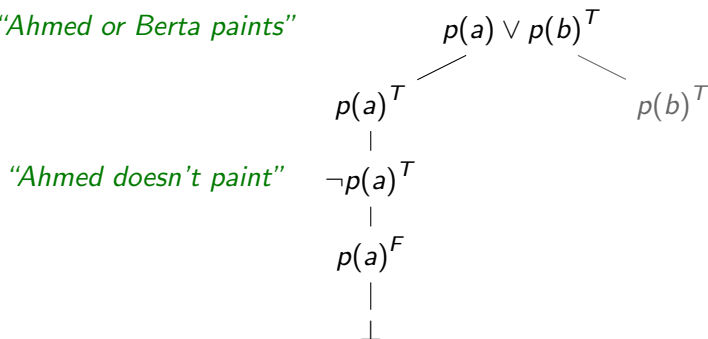
“Ahmed or Berta paints”



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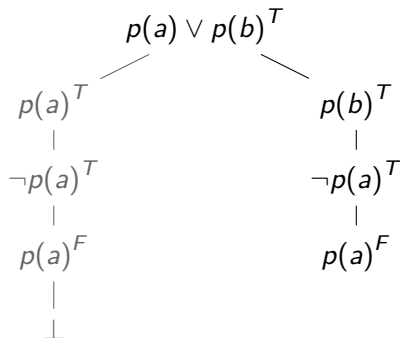
“Ahmed doesn't paint”

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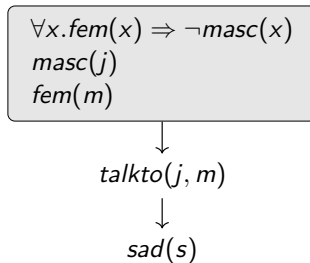
Background Knowledge

"John talks to Mary."

talkto(j, m)

"Sasha is sad."

sad(s)



Example: Tableaux Machine

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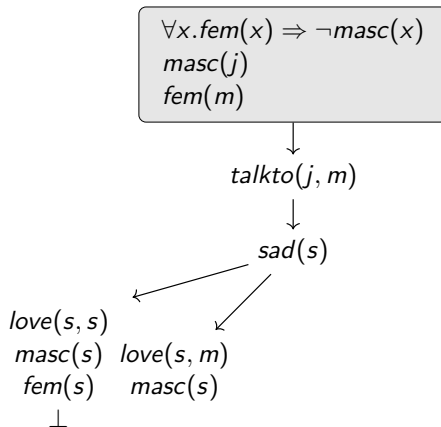
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$sad(s)$

"He loves her."

$\exists X.masc(X) \wedge$

$\exists Y.fem(Y) \wedge love(X, Y)$



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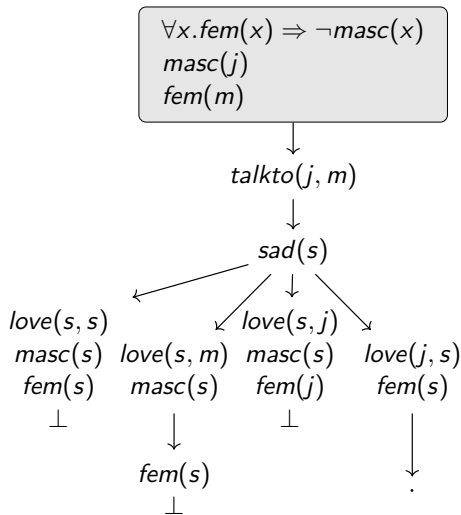
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"Sasha is a woman."

$fem(s)$



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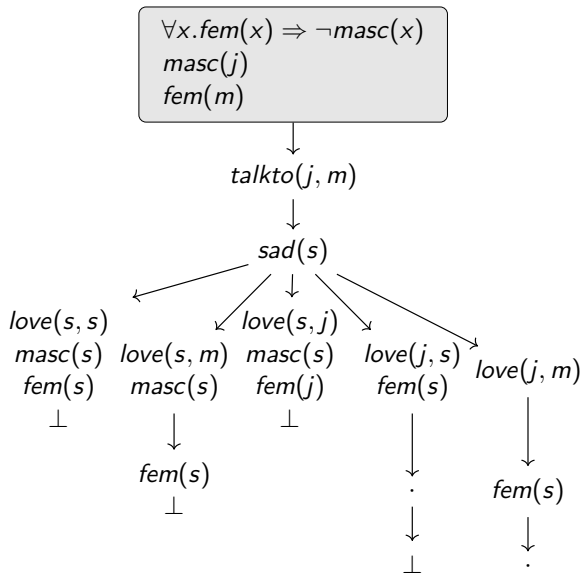
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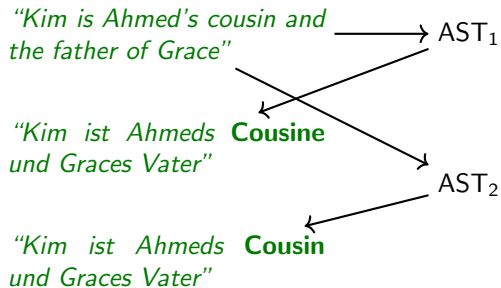
"John doesn't love Sasha."

$\neg love(j, s)$



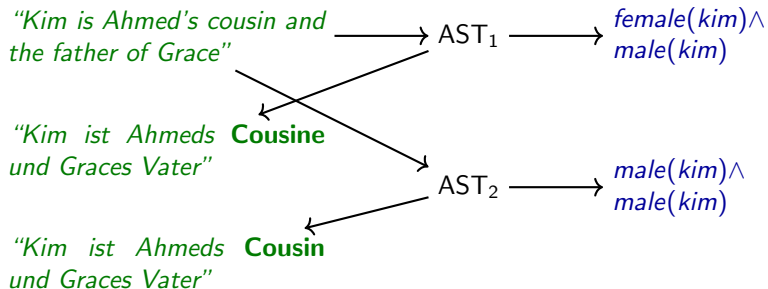
Example: Translation

- Two German words for “*cousin*”, depending on the gender
- Two entries in abstract syntax: `cousin_female` and `cousin_male`
- Use inference to discard ASTs



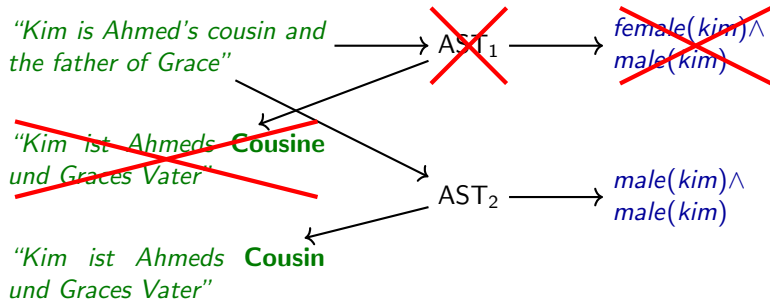
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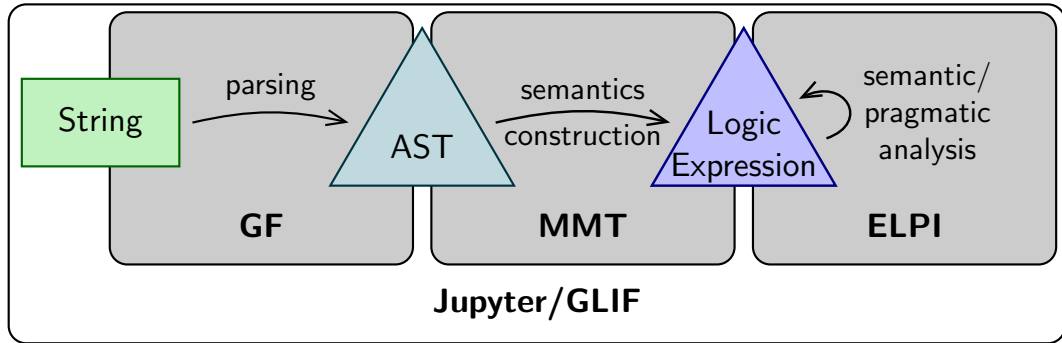


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GLIF Summary



Supporting the Semantic/Pragmatic Analysis

Observation:

- Parsing and semantics construction are based on specialized frameworks
- ELPI is a more general programming language

Can we do something more specialized?

- Not really – there is no “standard recipe” for semantic/pragmatic analysis
- But: It usually involves inferential reasoning

Let's generate provers!

Natural Deduction in MMT/LF

$$\frac{A \wedge B}{A} \wedge E$$

$$\frac{A \vee B \quad \begin{array}{c} [A]^1 \\ \vdots \\ C \end{array} \quad \begin{array}{c} [B]^1 \\ \vdots \\ C \end{array}}{C} \vee E^1$$

// $\vdash X$ is type of proofs for X (judgments as types)

$\vdash : o \rightarrow type$

$\wedge E1 : \Pi_{A:o} \Pi_{B:o} \vdash A \wedge B \rightarrow \vdash A$

$\vee E : \Pi_{A:o} \Pi_{B:o} \Pi_{C:o} \vdash A \vee B \rightarrow (\vdash A \rightarrow \vdash C) \rightarrow (\vdash B \rightarrow \vdash C) \rightarrow \vdash C$

Generating Provers in ELPI

LF rule $\wedge E1 : \Pi_{A:o} \Pi_{B:o} \vdash A \wedge B \rightarrow \vdash A$

ELPI equivalent

direct: `pi A \ pi B \ ded (and A B) => ded A.`

syn. sugar: `ded A :- ded (and A B).`

Generating Provers in ELPI

LF rule $\wedge E1 : \Pi_{A:o} \Pi_{B:o} \vdash A \wedge B \rightarrow \vdash A$

ELPI equivalent

direct: `pi A \ pi B \ ded (and A B) => ded A.`

syn. sugar: `ded A :- ded (and A B).`

Example: Or-Elimination

LF: $\vee E : \Pi_{A:o} \Pi_{B:o} \Pi_{C:o} \vdash A \vee B \rightarrow (\vdash A \rightarrow \vdash C) \rightarrow (\vdash B \rightarrow \vdash C) \rightarrow \vdash C$

ELPI: `ded C :- ded (or A B), ded A => ded C, ded B => ded C.`

Example: Forall-Introduction

LF: $\forall I : \Pi_{P:t \rightarrow o} (\Pi_{x:t} \vdash P \ x) \rightarrow \vdash \forall P$

ELPI: `ded (forall P) :- pi x \ ded (P x).`

Controlling the Proof Search

- Problem: Search diverges *searching harder than checking*
- Solution: Control search with helper predicates:
inspired by ProofCert project by Miller et al.
 - Intuition: Decide whether to apply rule
 - Do not affect correctness
 - Extra argument tracks aspects of proof state

Before: `ded A :- ded (and A B) .`

Now: `ded X A :- help/andEl X A B X1, ded X1 (and A B) .`

Helper Predicates

Name	Predicate	Argument
Iter. deepening	checks depth	remaining depth
Proof term	generates term	proof term
Product	calls other predicates	arguments for other predicates
Backchaining	Prolog's backchaining (\approx forward reasoning from axioms via \Rightarrow/\forall elimination rules)	pattern of formula to be proven (e.g. a conjunction)

Example helper: Iterative deepening

```
help/andEl (idcert  $N$ ) _ _ (idcert  $N1$ ) :-  $N > 0$ ,  $N1$  is  $N - 1$ .
```

Tableau Provers

$$\frac{A \wedge B^F}{A^F \mid B^F} \wedge^F \qquad \frac{A \wedge B^F \quad \begin{array}{c} [A^F] \\ \vdots \\ \perp \end{array} \quad \begin{array}{c} [B^F] \\ \vdots \\ \perp \end{array}}{\perp} \wedge^F$$

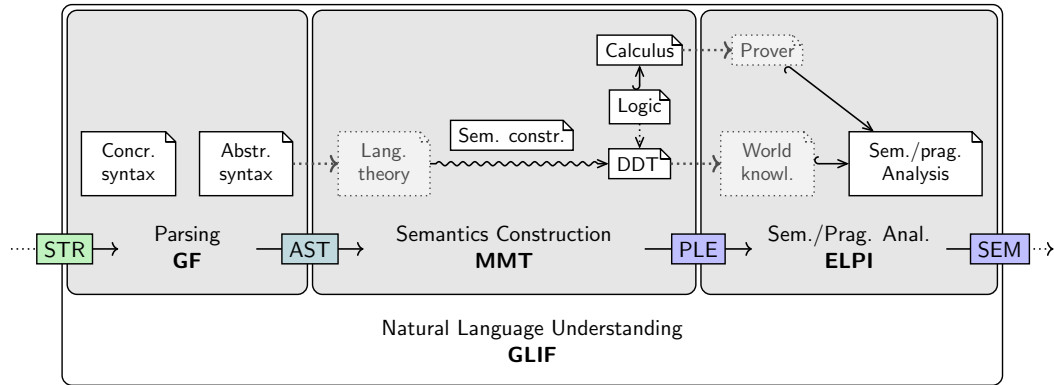
LF: $\wedge^F : \Pi_{A:o} \Pi_{B:o} A \wedge B^F \rightarrow (A^F \rightarrow \perp) \rightarrow (B^F \rightarrow \perp) \rightarrow \perp$

ELPI: `closed X :- help/andF X A B X1 X2 X3, f X1 (and A B),
f/hyp A => closed X2, f/hyp B => closed X3.`

With iterative deepening we get a working prover!

→ Other helpers result in more efficient provers

Conclusion: Prototyping NLU Pipelines



References I

- [GF] *GF - Grammatical Framework*. URL: <http://www.grammaticalframework.org> (visited on 09/27/2017).
- [KK03] Michael Kohlhase and Alexander Koller. “Resource-Adaptive Model Generation as a Performance Model”. In: *Logic Journal of the IGPL* 11.4 (2003), pp. 435–456. URL: <http://jigpal.oxfordjournals.org/cgi/content/abstract/11/4/435>.
- [LDM12] Hector Levesque, Ernest Davis, and Leora Morgenstern. “The Winograd Schema Challenge”. In: *Thirteenth International Conference on the Principles of Knowledge Representation and Reasoning*. 2012.
- [Mon70] R. Montague. “English as a Formal Language”. In: Reprinted in [Tho74], 188–221. Edizioni di Comunita, Milan, 1970, pp. 189–224.

References II

- [Mon74] Richard Montague. “The Proper Treatment of Quantification in Ordinary English”. In: *Formal Philosophy. Selected Papers*. Ed. by R. Thomason. New Haven: Yale University Press, 1974.
- [Tho74] R. Thomason, ed. *Formal Philosophy: selected Papers of Richard Montague*. Yale University Press, New Haven, CT, 1974.