Language Research in the KWARC group

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Workshop: Approaches to the Logic and Syntax of Mathematical Texts

Erlangen

Dec. 6, 2022

arxiv.org:

- Open-access pre-print server
- > 2,000,000 scientific articles
- Fields: physics, mathematics, computer science, ...
- LATEX sources
- → a great corpus

Problem:

```
This: "The average is \frac{A+B}{2}."
```

Could be written like this: The average is $\frac{A+B}{2}$.

Or like this: $\def \avg #1#2{\ensuremath {\frac {#1+#2}2}}$

The average is \ag{AB} .

with $Z(\beta, \alpha, \underline{\lambda})$ a normalization constant. Plugging this solution into the expression (C4) of G we get:

$$G(\underline{p},\underline{\lambda}) = \sum_{i=1}^{L} \sum_{y} \lambda_{i}(y) p_{i}(y) + \sum_{i=1}^{L-1} \sum_{y,y'} \lambda_{i,i+1}(y,y') p_{i,i+1}(y,y') - \frac{1}{\beta} \log Z(\beta,\alpha,\underline{\lambda}) . \tag{C7}$$

The condition on the Lagrange multipliers is finally obtained by looking at the stationary points of G with respect to the $\lambda_i(y)$'s, $\lambda_{i,i+1}(y,y')$'s. Let $\underline{\lambda}^*(\alpha)$ be a set of Lagrange multipliers achieving the stationary point, we then have

$$G(p) = G(p, \underline{\lambda}^*(\alpha))$$
,

where we have emphasized the dependence in α of the Lagrange multipliers.

Example from [BP22]

Problem:

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Or like this: $\def \avg#1#2{\ensuremath{\frac{#1+#2}2}}$

The average is \avg AB.

Solution: Convert to more managable format: HTML with MathML.

HTML, MathML

```
"The average is \frac{A+B}{2}."  
The average is <math>...</math>../p>
```

Presentation MathML

Content MathML

ar5iv corpus

Use LaTeXML to convert arxiv to HTML+MathML

Done by Deyan Ginev $\approx 2 \cdot 10^6$ documents

- → ar5iv corpus
- Goal: Extract semantic information and provide services

search, interactive documents, ...

Equation (12) looses validity as soon as target deformations start to become significant. The validity also depends on the accuracy of the mean longitudinal momentum given as a function of intensity. For $I\lambda^2 = 1.0 \cdot 10^{17} \text{Wcm}^{-2} \mu\text{m}^2$ we obtain an ejection angle of $\theta' = 14^{\circ}$ and for Highlight annotations $I\lambda^2 = 2.0 \cdot 10^{18} \text{Wcm}^{-2} \mu \text{m}$ mulations. This vields $\alpha^{-1} \approx 8.0 \cdot 10^{17} \text{W}$ Convert all to basic SI units watt Watt In conclusion, we have centimeter^-2 horsepower simulation techniques micrometer^2 L sun can be emitted from a corona is present. In a Reset this at electrons are injected into the overn and injection Reset Document directions are almost a

Screenshot from [MK]

ar5iv corpus

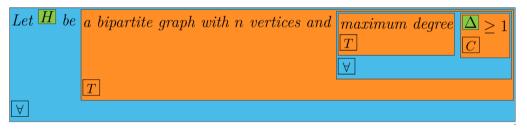
Use LaTeXML to convert arxiv to HTML+MathML

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→ ar5iv corpus

- $pprox 2 \cdot 10^6$ documents
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search, interactive documents, ...



Screenshot from [Sch16]

—And Now for Something Completely Different—

GLIF: A tool for prototyping natural language semantics

Natural Language Semantics (Symbolic)

For me:

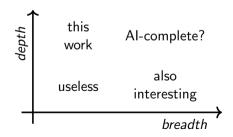
Translating natural language into a formal semantic representation (logic).

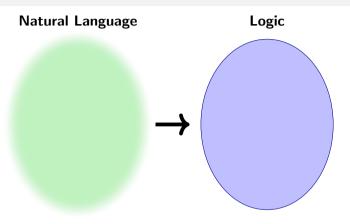
Example:

"Every student paints and is quiet." $\rightsquigarrow \forall x.s(x) \Rightarrow (p(x) \land q(x))$

Rule-based (no ML):

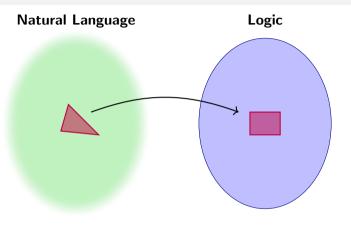
Parsing → semantics construction → inference.





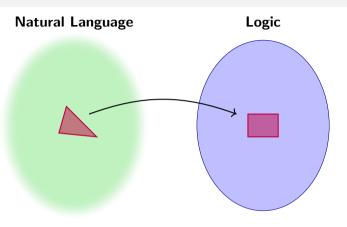
How do we get from messy language to formal logic?

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



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

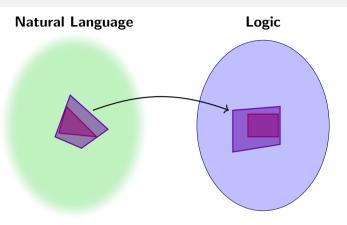


"Ahmed paints and Berta is quiet."

"Ahmed doesn't paint."

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

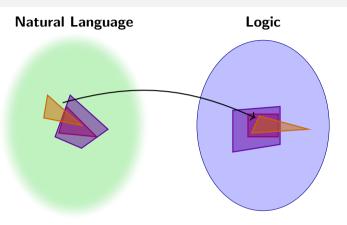
 $\neg p(a)$



"Every student paints and is quiet."

"Nobody paints."

$$\forall x.s(x) \Rightarrow (p(x) \land q(x))$$
$$\neg \exists x.p(x)$$



"Ahmed isn't allowed to paint."

"Ahmed and Berta must paint."

$$\neg \Diamond p(a)$$

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

Hand-waving is problematic:

"Ahmed paints. He is quiet." $\stackrel{!}{\leadsto}$ $p(a) \land q(a)$

Montague: Specify

grammar,

fixes NL subset

- target logic,
- semantics construction.

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 $[\phi \land \psi]$, ϕ or ψ translates into $[\phi \lor \psi]$.
- T12. If $\gamma, \delta \in P_{IV}$ and γ, δ translate into γ', δ' respectively, then γ and δ translates into $\hat{x}[\gamma'(x) \wedge \delta'(x)], \gamma$ 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 $\widehat{P}[\alpha'(P) \vee \beta'(P)]$.

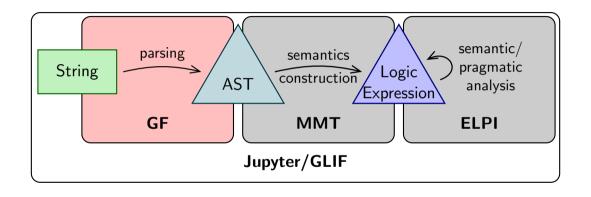
Claim: That doesn't scale well → We need prototyping!

NLU Prototyping

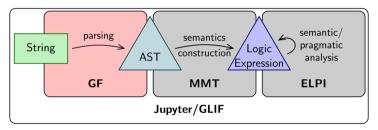
```
> translate "Every student paints and is quiet." \forall x.s(x) \Rightarrow (p(x) \land q(x)) > answer "Every student is quiet. John is a student. Is John quiet?" \forall x.s(x) \Rightarrow q(x), s(j) \vdash^{?} q(j) yes
```

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

Components of GLIF: GF

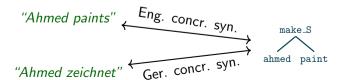


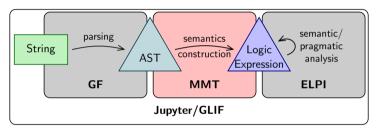
Components of GLIF: GF



Components of GLIF: Grammatical Framework [GF]

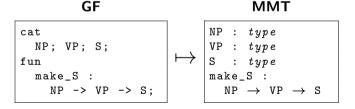
- Specialized for developing natural language grammars
- Separates abstract and concrete syntax
 make_S : NP -> VP -> S;
 abstract
 make_S np vp = np.s ++ vp.s!np.n;
 concrete
- Abstract syntax based on LF
- Comes with large library ≥ 36 languages





- Modular logic development and knowledge repr.
- Not specialized in one logical framework we use LF
- We will use MMT to:
 - 1 represent abstract syntax
 - 2 specify target logic and discourse domain theory
 - 3 specify semantics construction

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Logic Syntax

```
o: type //propositions
\neg: o \rightarrow o
\land: o \rightarrow o \rightarrow o
\lor: o \rightarrow o \rightarrow o
\iota: type //individuals
\forall: (\iota \rightarrow o) \rightarrow o
\exists: (\iota \rightarrow o) \rightarrow o
```

Discourse Domain

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

- Modular logic development and knowledge repr.
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- We will use MMT to:
 - 1 represent abstract syntax
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Semantics Construction

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

Simple setting

More advanced

Example: Parsing + Semantics Construction

"Ahmed and Berta paint"

↓parsing

make_S (andNP ahmed berta) paint

↓semantics construction

 $(\lambda n. \lambda v. n\ v)\ ((\lambda a. \lambda b. \lambda p. a\ p\ \wedge\ b\ p)\ (\lambda p. p\ ahmed)\ (\lambda p. p\ berta))\ pair$

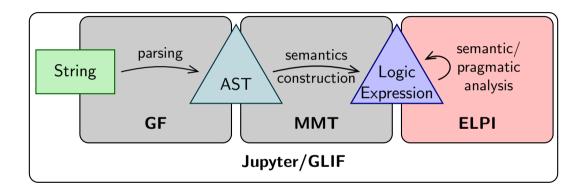
 \downarrow_{β} -reduction

paint ahmed ∧ paint berta

Example: Student Project [Int]

```
parse "John has not always run" | construct
¬H (run iohn)
parse "John has to have been allowed to always run" | construct
\Box P \diamond (H (run iohn) \land G (run iohn))
parse "John probably will never run" | construct
Prob G ¬(run john)
parse "it has to be possible that John runs" | construct
□ ⟨run john⟩
parse "Mary saw that John would kill the dog" | construct
P ∞marv∞F (kill john dog)
parse "Mary runs and John sees it" | construct
(run marv)∧⊛iohn⊛(run marv)
```

Components of GLIF: ELPI



Components of GLIF: ELPI

• Implementation and extension of $\lambda Prolog$

 \approx *Prolog* + *HOAS*

- MMT can generate logic signatures
- First experiments with prover generation
- Generic inference/reasoning step after semantics construction

MMT

ELPI

"the ball has a mass of 5kg" \rightarrow AST \longrightarrow mass(theball, quant(5, kilo gram))

"the ball has a mass of
$$5kg$$
" \longrightarrow AST \longrightarrow mass(theball, quant(5, kilo gram))

AST₁ $\longrightarrow \lambda x. E_{kin}(x, quant(2, milli Newton))$

"a kinetic energy of $12mN$ "

AST₂ $\longrightarrow \lambda x. E_{kin}(x, quant(2, meter \cdot Newton))$

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"a kinetic energy of $12mN$ "

AST₂ $\longrightarrow \lambda x. E_{kin}(x, quant(2, meter Newton))$

```
In [20]: 

| parse "the ball has a mass of 5 k g and a kinetic energy of 12 m N" |
| (mass theball (quant 5 kilo gram)) ∧ (ekin theball (quant 12 milli Newton))
| (mass theball (quant 5 kilo gram)) ∧ (ekin theball (quant 12 meter·Newton))

In [21]: 
| parse "the ball has a mass of 5 k g and a kinetic energy of 12 m N" |
| construct | filter -predicate=filter_pred
| (mass theball (quant 5 kilo gram)) ∧ (ekin theball (quant 12 meter·Newton))
```

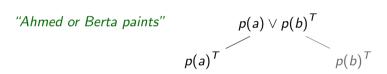
Example: Epistemic Q&A

```
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)
```

$$egin{aligned} S_1 &= \Box_{john}(\Box_{mary}hd(ping) ee \Box_{eve}hd(ping)) \ S_2 &= \neg(\Box_{mary}hd(ping) ee \Box_{mary}\neg hd(ping)) \ Q &= \Box_{eve}hd(ping) ee \Box_{eve}\neg hd(ping) \end{aligned}$$

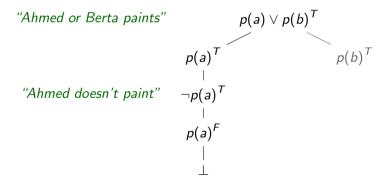
Example: Tableaux Machine [KK03]

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



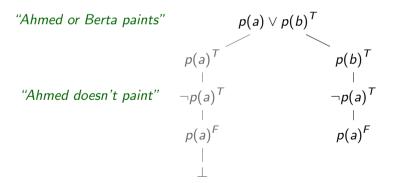
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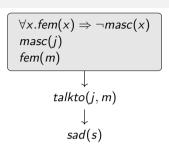
${\sf Background}\ {\sf Knowledge}$

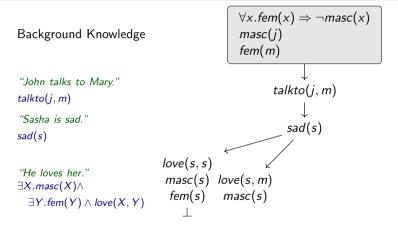
```
"John talks to Mary."

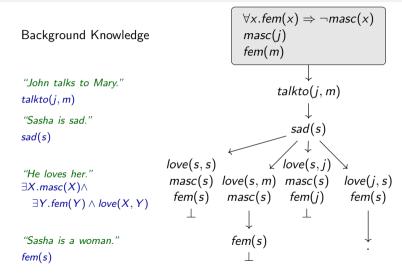
talkto(j, m)

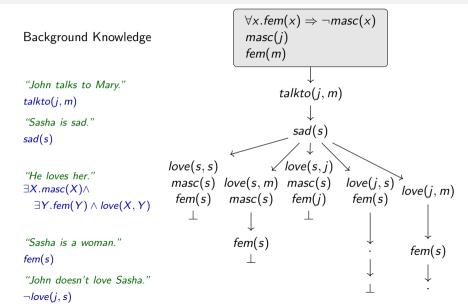
"Sasha is sad."

sad(s)
```









Example: Input Language for SageMath

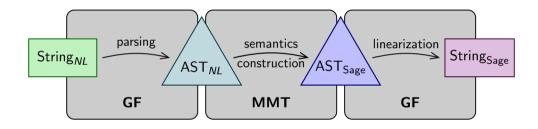
• Can we make a natural input language for SageMath?

WolframAlpha-like

```
sage: g = AlternatingGroup(5)
sage: g.cardinality()
60
```

"Let G be the alternating group on 5 symbols. What is the cardinality of G?"

Example: Input Language for SageMath



Example: Input Language for SageMath

```
> Let G be the alternating group on 5 symbols.
# G = AlternatingGroup(5)
> Let |H| be a notation for the cardinality of H.
# def bars(H): return H.cardinality()
> What is |G|?
# print(bars(G))
60
> Let A<sub>N</sub> be a notation for the alternating group on N symbols.
# def A(N): return AlternatingGroup(N)
> What are the cardinalities of A_4 and A_5?
# print(A(4).cardinality()); print(A(5).cardinality())
12
60
```

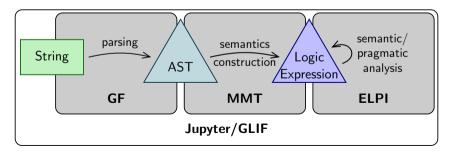
Conclusion

Summary:

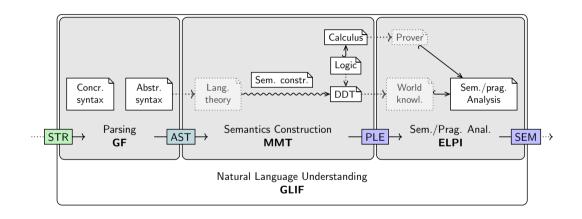
- GLIF = GF + MMT + ELPI
- Prototyping natural language semantics
- We use it for teaching

Examples:

- 1 "a kinetic energy of 12mN"
- "He loves her" (tableaux machine)
- 3 "John knows that Eve has a dog"
- 4 "What is the cardinality of G?"



Pipeline Specification



References I

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- [GF] GF Grammatical Framework. URL: http://www.grammaticalframework.org (visited on 09/27/2017).
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- [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.

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