

Textual Entailment: Bridging Logic and Language

Valeria de Paiva

Nuance Communications, NL and AI Lab, Sunnyvale,
Visiting Prof CSF, DI-PUC-RJ



Nuance Comms, AI and NL Lab, Sunnyvale, CA



+

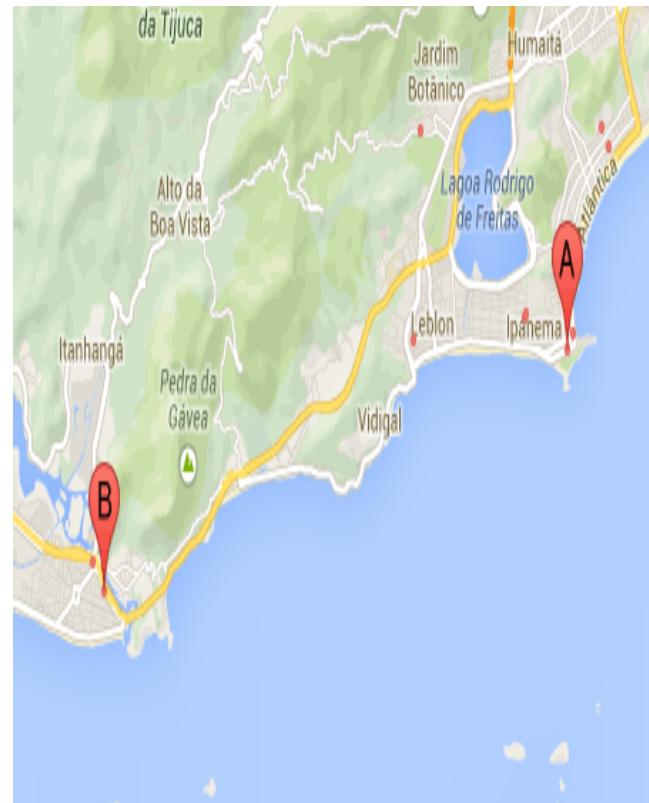
Ron Kaplan “**Beyond the GUI: It’s Time for a Conversational User Interface**”, Wired 2013



<http://www.wired.com/2013/03/conversational-user-interface/>

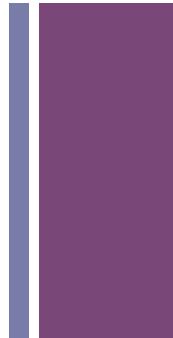
+

Siri and other personal assistants



+

The Future is Meaning...



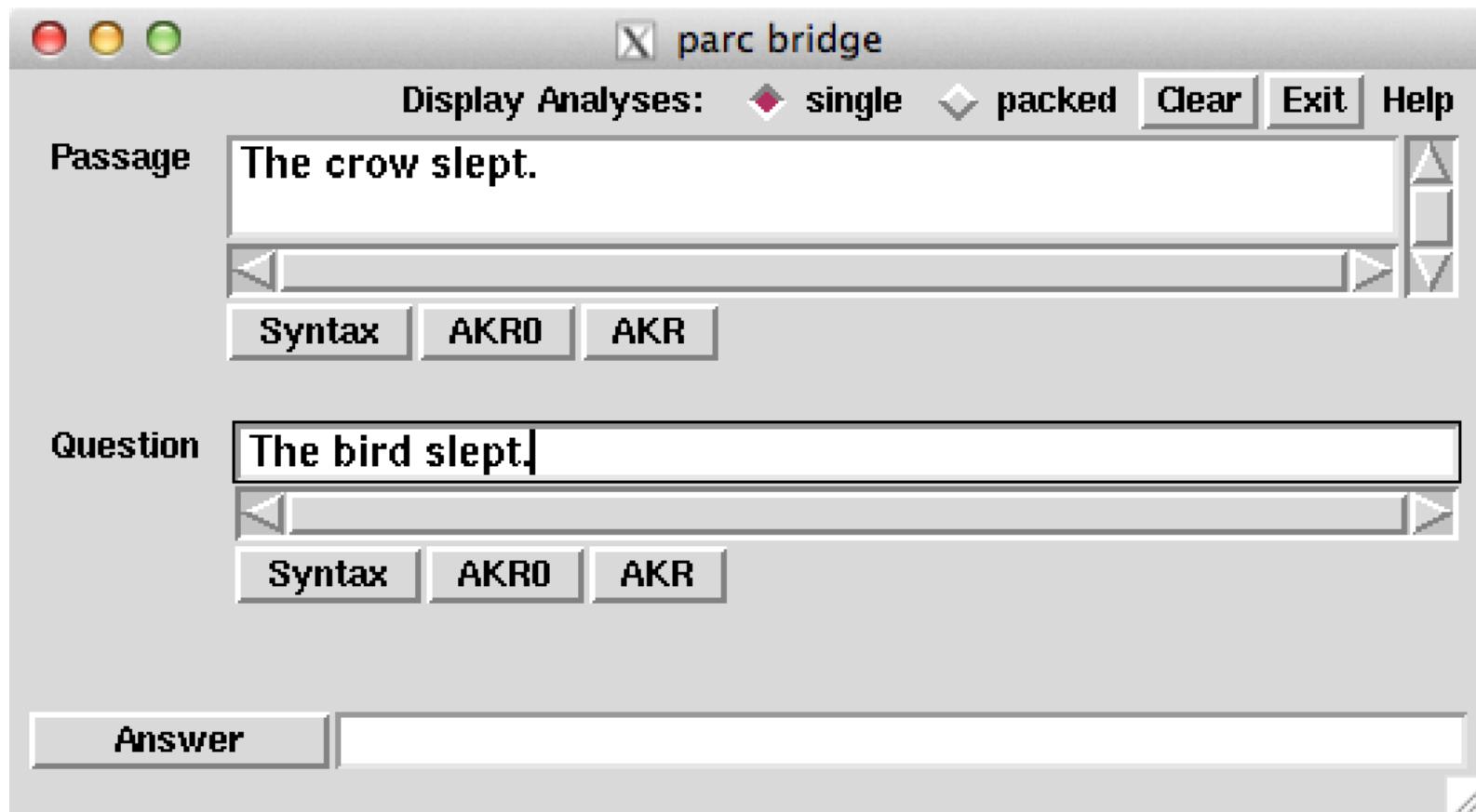
+

Recent Past...

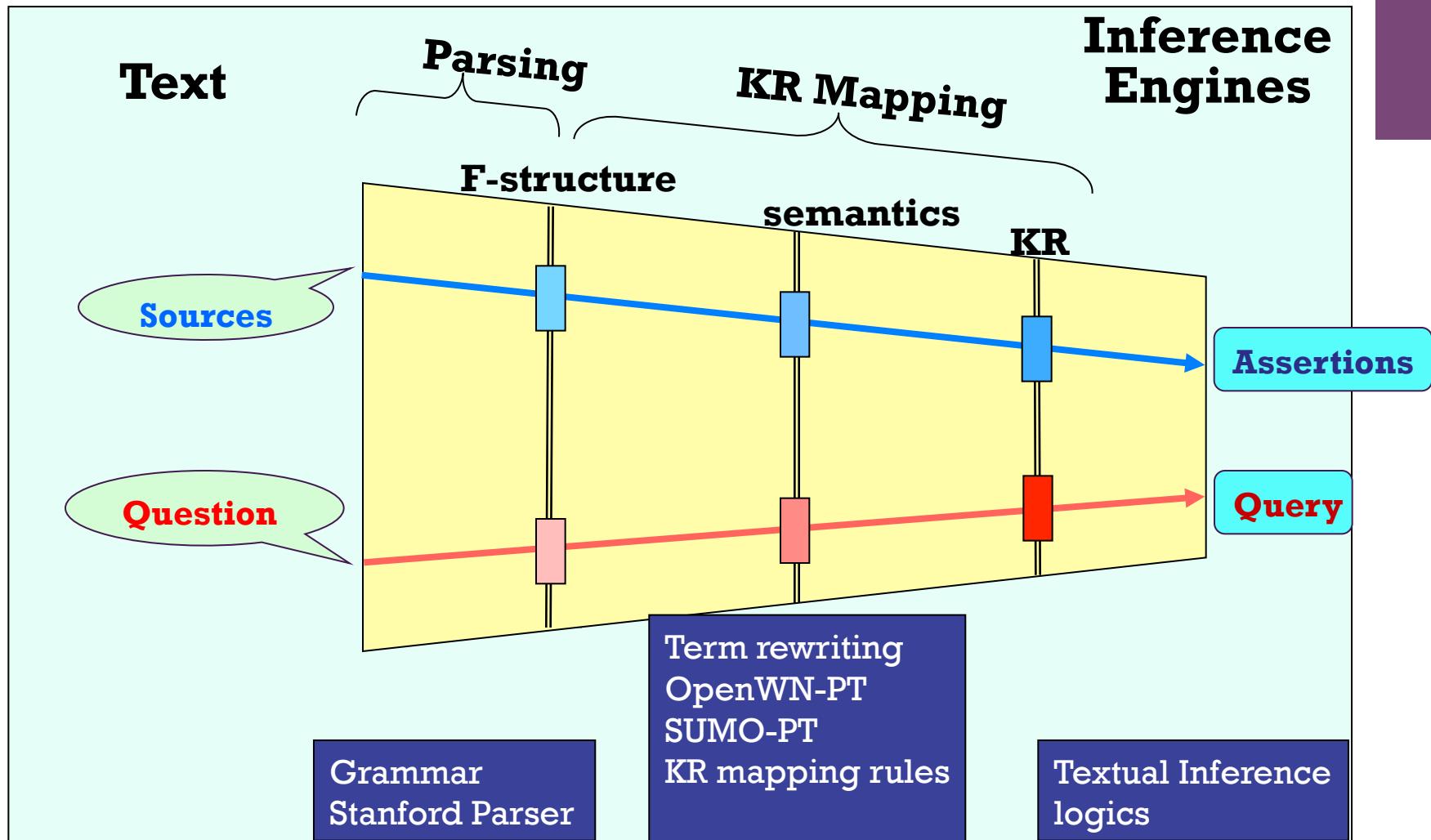




PARC's Bridge System (1999-2008)



+ PARC's Bridge System (1999-2008)

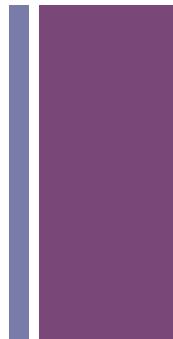




Powerset

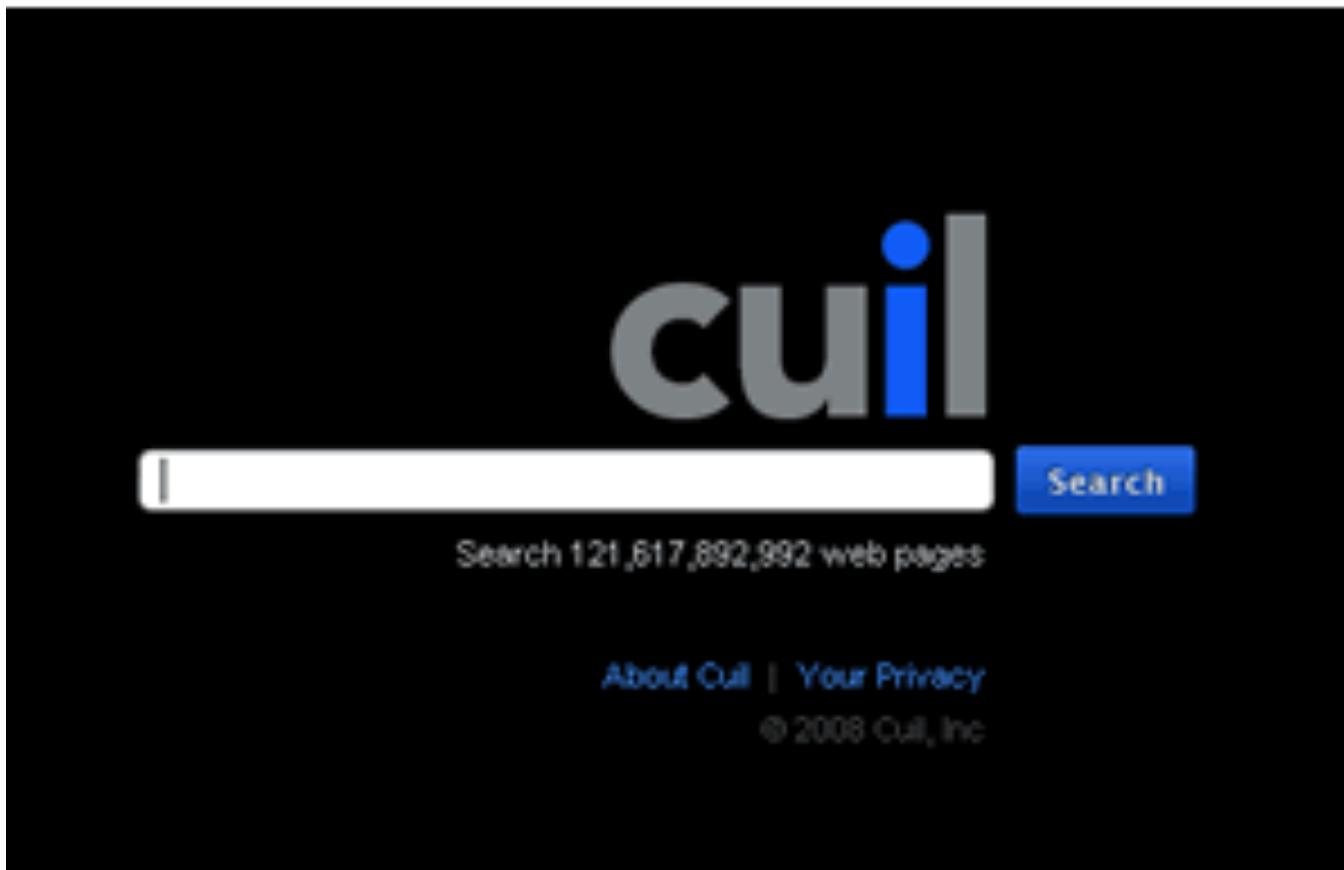
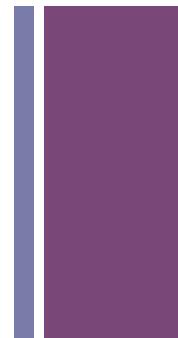


Acquired by Microsoft, 2008



+

and Cuil...



The image shows the homepage of the Cuil search engine. The background is black. At the top center is the Cuil logo, where the letters 'c' and 'u' are in grey and 'i' is in blue. Below the logo is a search bar with a white input field containing a single vertical bar character and a blue 'Search' button to its right. Underneath the search bar, the text 'Search 121,617,892,992 web pages' is displayed. At the bottom of the page, there are links for 'About Cuil | Your Privacy' and the copyright notice '© 2008 Cuil, Inc'.

cuil

|

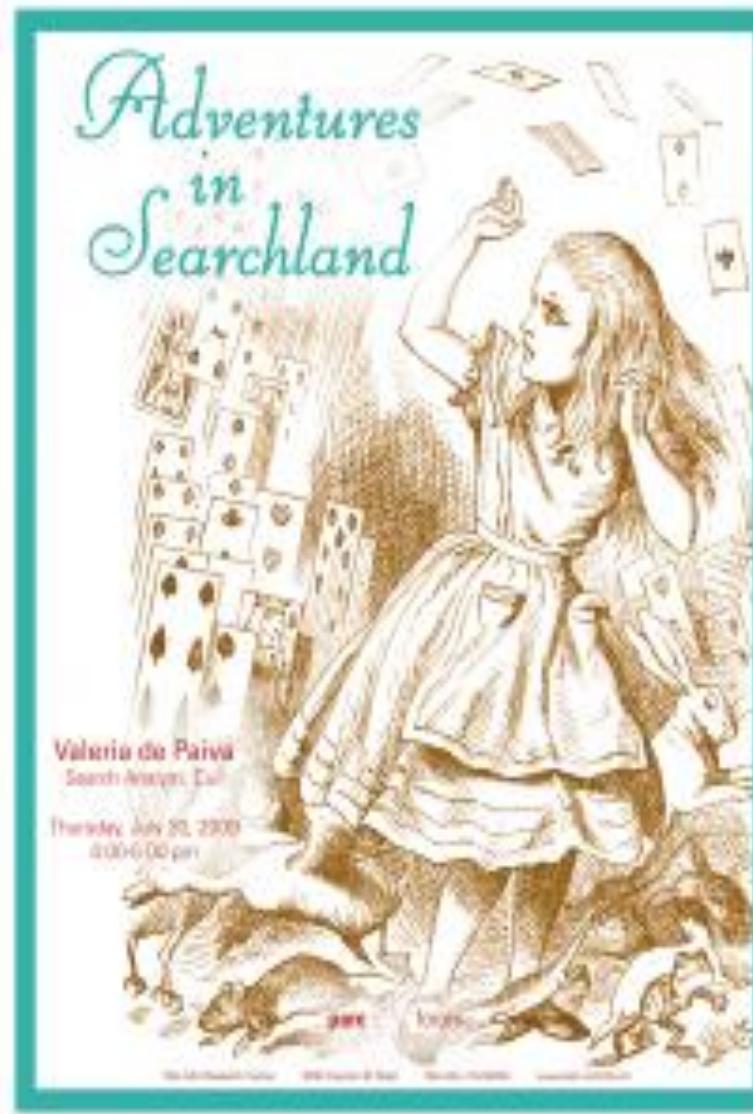
Search

Search 121,617,892,992 web pages

About Cuil | Your Privacy
© 2008 Cuil, Inc



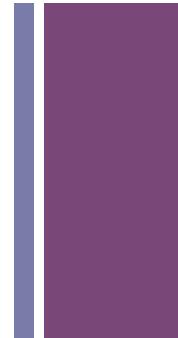
Another storv



[https://www.parc.com/event/934/adventures-in-searchland.html](http://www.parc.com/event/934/adventures-in-searchland.html)

+

Nowadays: redoing PARC work in Portuguese...

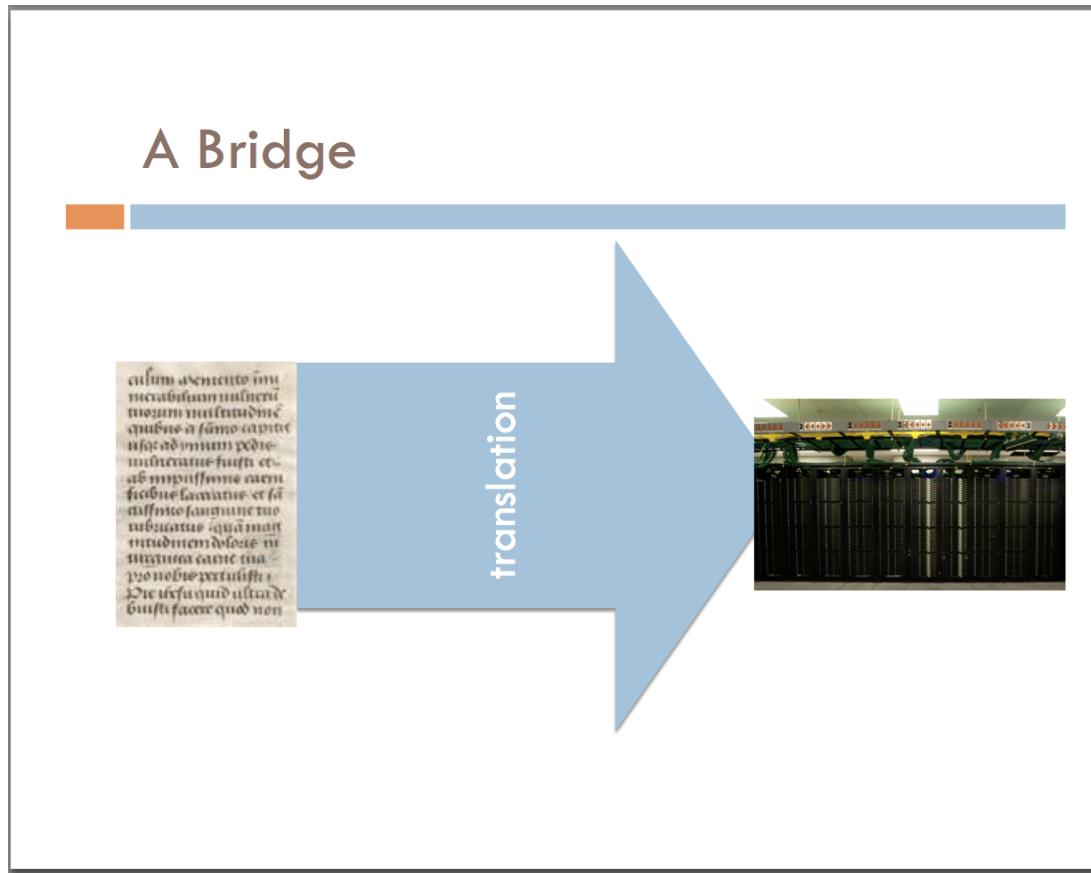


PROCESSAMENTO DE
LINGUAGEM NATURAL:
UMA PERSPECTIVA
PESSOAL WINGING NLP...

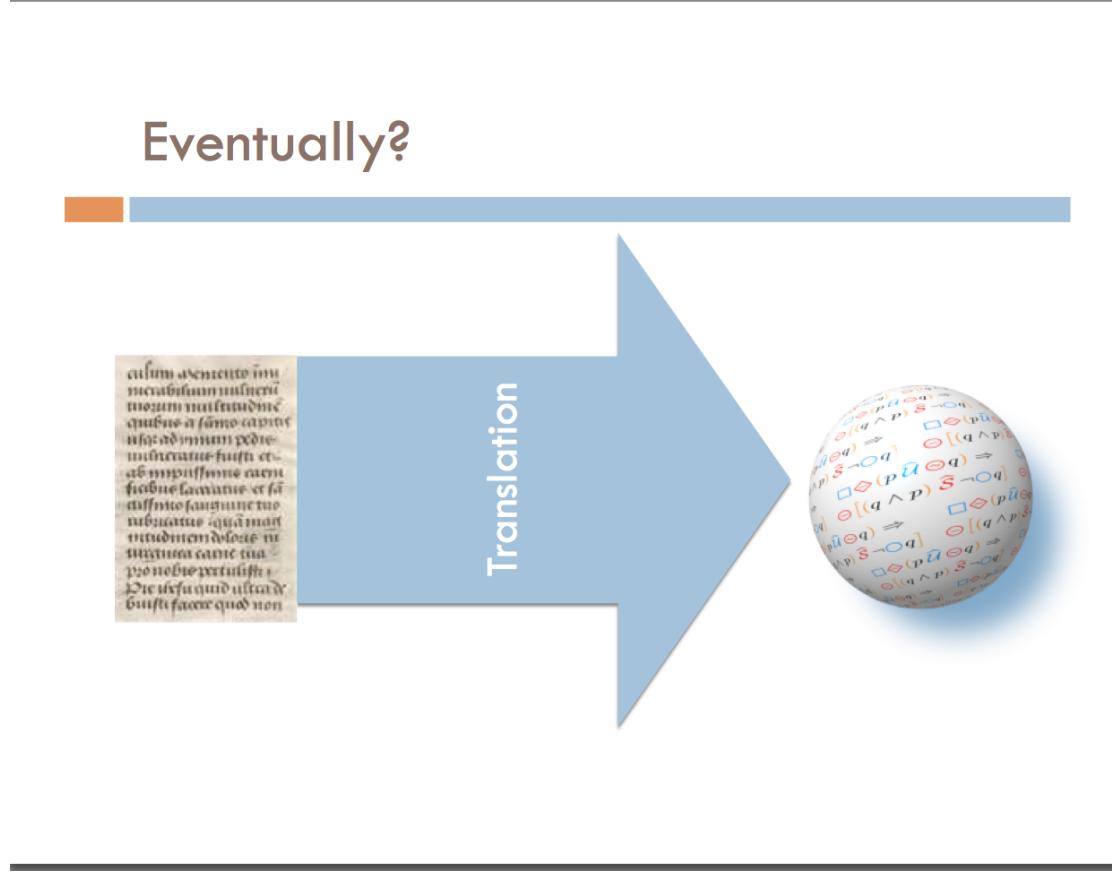
Valeria de Paiva, Cuil, Inc.
FGV Setembro de 2010



Goals in 2010



Goals in 2010





Goals in 2010...

In fact we want...

?

caſum aveniente inu
incabiliu[m] inibueri
meum inſtanciam
quibus a ſamo capiat
ut ad omnia pede-
mbeatur fuſt et
ab impaſſione cu[m]
ſtobus lauantur et ſi
diffimmo ſanguinetu[m]
subducatur quādā
inuidem deſtitit in
meum canis ma-
rionebus retulit
Die uſtu[m] ultat de
buſſu facie quod non

Translation



- Content analysis /large-scale intelligent information extraction, access and retrieval
- Text understanding
- Text generation
- Text simplification
- Automatic summarization
- Dialogue systems
- Question answering
- Machine Translation
- Named Entity Recognition,
- Anaphora/co-reference resolution,
- Reading, writing, grammar aids, etc...

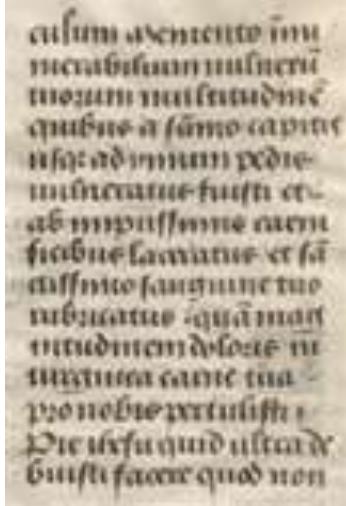


Goals in 2014...

- The same!
- But we've done LOTS...
- Only in 2014, more than 9 papers (5 to be presented next month, in DHandES, PROPOR and TorPorEsp) on systems for Portuguese
- TO RECAP:

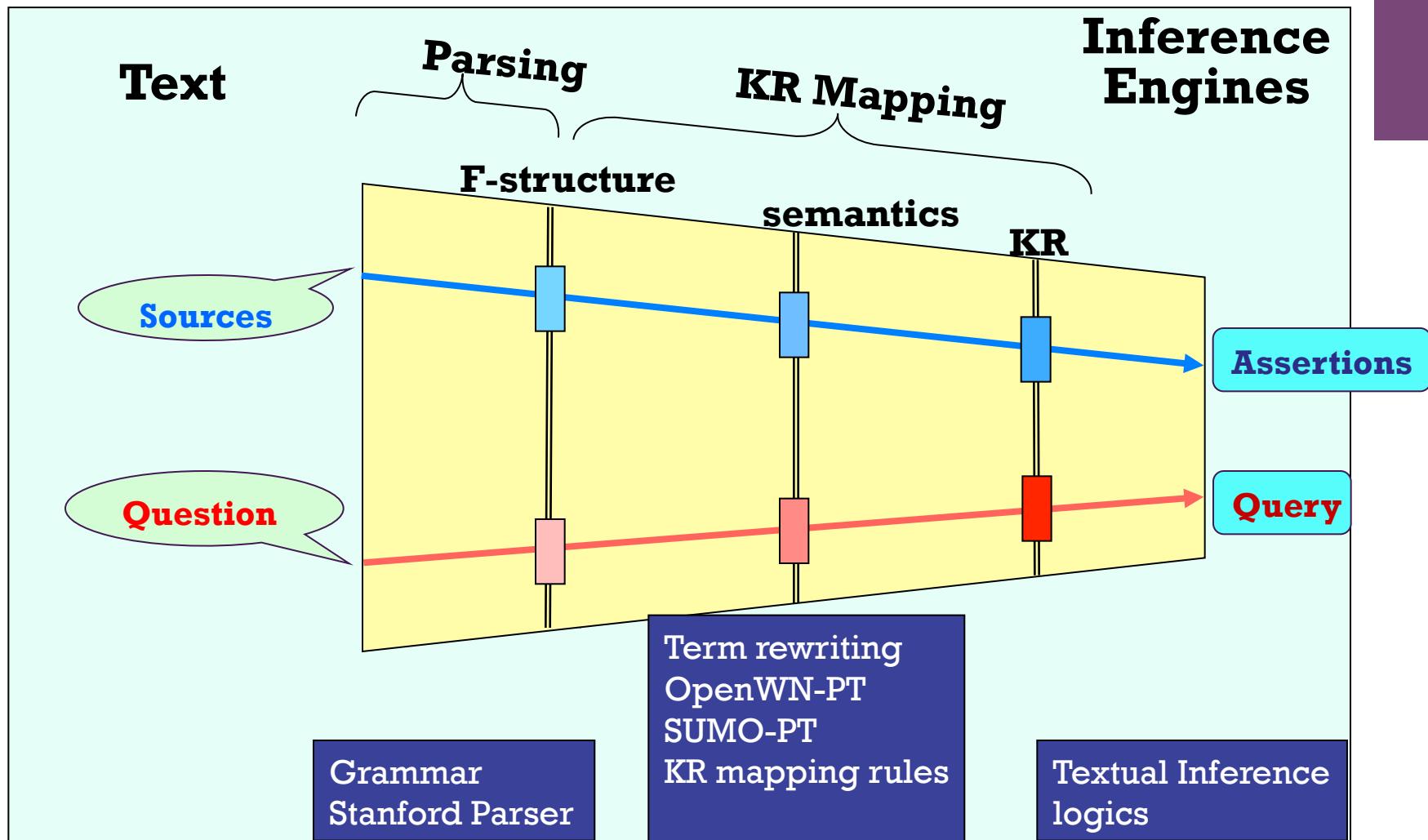
+ What can we do? Logic and Lexical Ontologies

Group: Alexandre Rademaker, Bruno Lopes,
Claudia Freitas, Dario Oliveira, Gerard de Melo,
Livy Real, Suemi Higuchi, Hermann Hauesler, Luiz
Carlos Pereira, Vivek Nigam and Valeria de Paiva



Improving Lexical Resources
and Inferential Systems to
work with Logic coming from
free form text.

+ PARC's Bridge System (1999-2008)



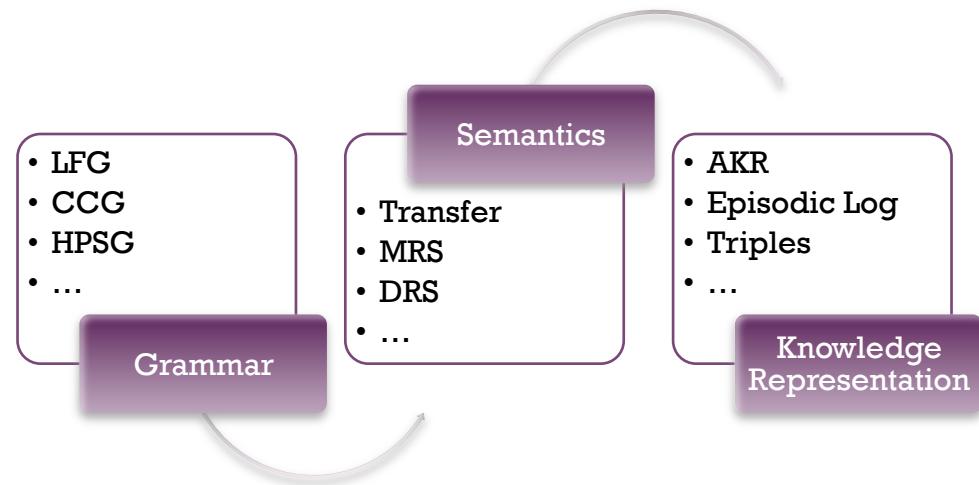
Idea: Simplify and reproduce components in PORTUGUESE



A Generic Architecture

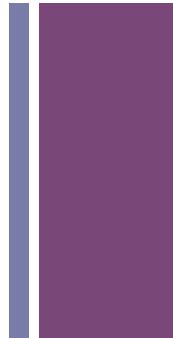
All require a host of pre & post-processing: text segmenters, POS taggers, Lexica, Named Entity Recognizers, Gazetteers, Temporal Modules, Coreference Resolution, WSD, etc

Pipeline Envisaged in 2010





Reality Check...

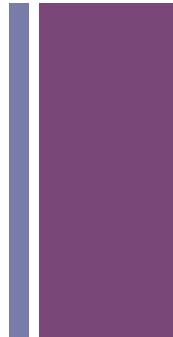


- What PARC considered pre-processing is MOST of the processing...
- Got the XLE research license, but hard to use it, needed several lexicons that DO NOT exist in Portuguese, notably WordNet
- There are several open toolkits that can be used instead:
FREELING
OpenNLP
StanfordNLP
NLTK

More usable, more community, less expertise required



Today: Inference, any one?...



- **Textual entailment** methods recognize, generate, and extract pairs $\langle T, H \rangle$ of natural language expressions, such that a human who reads (and trusts) T would infer that H is most likely also true (Dagan, Glickman & Magnini, 2006)
- Example:
(T) The drugs that slow down Alzheimer's disease work best the earlier you administer them.
(H) Alzheimer's disease can be slowed down using drugs.
 $T \Rightarrow H$
- A series of competitions since 2004, ACL "Textual Entailment Portal", many different systems...



RTE Competitions

- 15 meetings so far in
[http://aclweb.org/aclwiki/index.php?
title=Textual Entailment References](http://aclweb.org/aclwiki/index.php?title=Textual_Entailment_References)
- A BIG area, lots of research: tutorials, books, courses...
- [8th Recognizing Textual Entailment Challenge](#) at SemEval 2013
- [...] NIST and PASCAL
- [ACL 2005 Workshop on Empirical Modeling of Semantic Equivalence and Entailment, 2005](#)
[First PASCAL Recognising Textual Entailment Challenge \(RTE-1\), 2005](#)



All Logic?...

- By no means
- Mostly NO Logic...
- Graphs, alignments, transformations, stats
- Some logic though: Stanford
(MacCartney&Manning, Bos, etc..)
- Today: Inference using theorem proving
- Vivek Nigam, UF Paraiba





BlackBox Inference: outline

- Use Xerox's PARC Bridge system as a black box to produce NL representations of sentences in KIML (Knowledge Inference Management Language).
- KIML + inference rules = TIL (version of) Textual Inference Logic
- Translate TIL formulas to a theory in Maude, the SRI rewriting system.
- Use Maude rewriting to prove Textual Entailment “theorems”.



An example: a crow slept

■ Conceptual Structure:

```
role(cardinality restriction,crow-1,sg)  
role(sb,sleep-4,crow-1)  
subconcept(crow-1,  
[crow#n#1,crow#n#2,brag#n#1])  
subconcept(sleep-4,  
[sleep#v#1,sleep#v#2])
```

■ Contextual Structure:

```
instantiable(crow-1,t)  
instantiable(sleep-4,t)
```

top context(t)

Temporal Structure:

```
trole(when,sleep-4,interval(before,Now)  
)
```





KIML (Knowledge Inference Management Language)

- A representation language based on events (neo-Davidsonian), concepts, roles and contexts, McCarthy-style
- Using events, concepts and roles is traditional in NL semantics
- Usually equivalent to FOL (first-order logic), ours a small extension, contexts are like **modalities**.
Language based on linguists' intuitions !
- Exact formulation still being decided: e.g. not considering temporal assertions, yet...





KIML versus FOL

- In FOL could write $\exists \text{Crow} \exists \text{Sleep}.\text{Sleep}(\text{crow})$
Instead we will use basic concepts from a parameter ontology
- O (could be Cyc, SUMO, UL, KM, etc...)
- Instead of FOL have Skolem constant crow-1 a subconcept of an ambiguous list of concepts:
`subconcept(crow-1,
[crow#n#1,crow#n#2,brag#n#1])`
- Same for sleep-2 and have roles relating concepts
`role(sb,sleep-4,crow-1)`
meaning that the sb=subject of the sleeping event is a crow concept

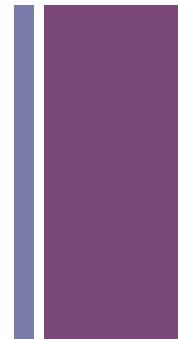


What is Different?

- Corresponding to formulas in FOL, KIML has a collection of assertions that, read conjunctively, correspond to the semantics of a (fragment of a) sentence in English.
- Concepts in KIML – similar to Description Logic concepts primitive concepts from an idealized version of the chosen
- Ontology on-the-fly concepts, always sub-concepts of some primitive concept. concepts are as fine or as coarse as needed by the application
- Roles connect concepts: deciding which roles with which concepts a big problem... for linguists
- Roles assigned in a consistent, coherent and maximally informative way by the NLP module



Contexts for Quantification



- Using contexts for modelling negation, implication, as well as propositional attitudes and other intensional phenomena. There is a first initial context (written as t), roughly what the author of the sentence takes the world to be.
- Contexts used for making existential statements about the existence and non-existence in specified possible worlds of entities that satisfy the intensional descriptions specified by our concepts. Propositional attitudes predicates (knowing, believing, saying,...) relate contexts and concepts in our logic.
- Concepts like knowing, believing, saying introduce context that represents the proposition that is known, believed or said.

+ Ed knows that the crow slept

- alias(Ed-0,[Ed])
role(prop,know-1,ctx(sleep-8))
role(sb,know-1,Ed-0)
role(sb,sleep-8,crow-6)
subconcept(Ed-0,[male#n#2])
subconcept(crow-6,
[crow#n#1,crow#n#2,brag#n#1])
subconcept(know-1,[know#v#1,...,sleep-
together#v#1]) subconcept(sleep-8,
[sleep#v#1,sleep#v#2]) context(ctx(sleep-8)),
context(t) context-lifting-
relation(veridical,t,ctx(sleep-8)) context-
relation(t,ctx(sleep-8),crel(prop,know-1))
instantiable(Ed-0,t)
instantiable(crow-6,ctx(sleep-8))
instantiable(sleep-8,ctx(sleep-8))

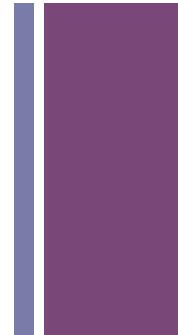


Inference to build reps and to reason with them

- In previous example can conclude:
`instantiable(sleep-8,t)`
if knowing X implies X is true.
(Can conclude `instantiable(crow-6,t)` too, for definitiveness reasons..)
- Happening or not of events is dealt with by the **instantiability/uninstantiability** predicate that relates concepts and contexts
e.g. Negotiations prevented a strike
- Contexts can be:
veridical, antiveridical or averidical with respect to other contexts.
- Have '**context lifting rules**' to move instantiability assertions between contexts.



Inference Rules



The totally obvious

$$\frac{}{s \rightarrow s}$$

$$\frac{s \rightarrow t \quad s \rightarrow r}{r \rightarrow t}$$



Inference Rules

Inheritance rules

Nina has a canary, canary \sqsubseteq bird

Ed kissed Nina, kiss \sqsubseteq touch

Nina has a bird

Ed touched Nina

Every carp is a fish, carp \sqsupseteq koi

Every koi is a fish

She didn't give him a bird, bird \sqsupseteq canary

She didn't give him a canary



Modifiers Inference Rules

Ed arrived in the city by bus

Ed did not arrive in the city

Ed arrived in the city

Ed did not arrive in the city by bus

Ed arrived in the city, Ed \sqsubseteq person

Ed arrived in Rome, Rome \sqsubseteq city

A person arrived in the city

Ed arrived in a city

Note that *Ed did not arrive in the city by bus* does **not** entail that
Ed did not arrive in the city.



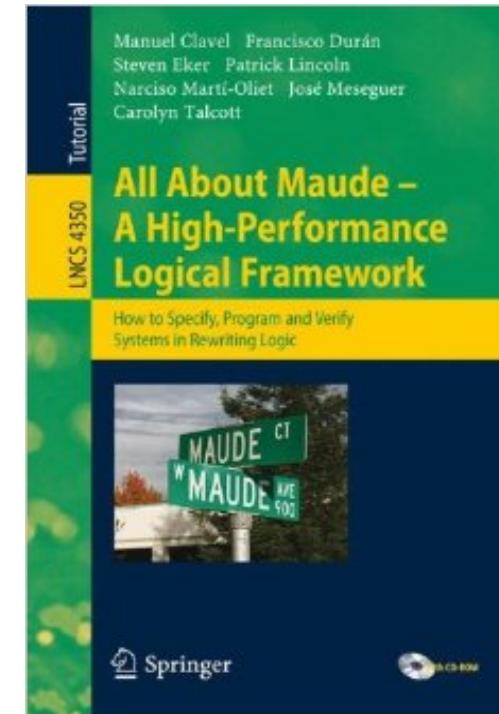
Implicative Commitment Rules

- Preserving polarity:
“Ed managed to close the door” → “Ed closed the door”
“Ed didn’t manage to close the door” → “Ed didn’t close the door”.
- The verb “forget (to)” inverts polarities:
“Ed forgot to close the door” → “Ed didn’t close the door”
“Ed didn’t forget to close the door” → “Ed closed the door”.
- There are six such classes, depending on whether positive environments are taken to positive or negative ones.
- Accommodating this fine-grained analysis into traditional logic description is further work. (Nairn et al 2006 presents an implemented recursive algorithm for composing these rules)



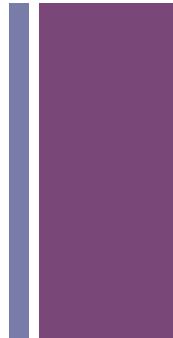
Towards a Rewriting Framework

- A implementation of TIL, using the traditional rewriting system Maude to reason about the logical representations produced by the blackbox NLP module
- Hand-correct the representations given by the NLP module: the goal here is not to obtain correct representations, but to work logically with correct representations.
- Maude system is an implementation of rewriting logic developed at SRI International.
- Maude modules (rewrite theories) consist of a term-language plus sets of equations and rewrite-rules. Terms in rewrite theory are constructed using operators (functions taking 0 or more arguments of some sort, which return a term of a specific sort).

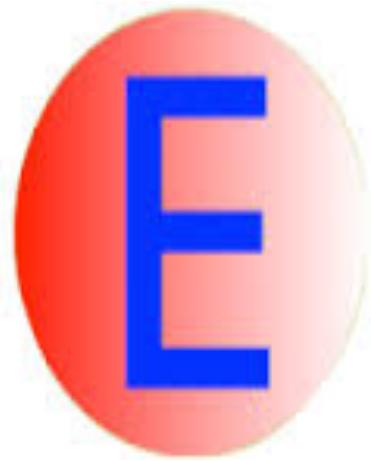




A Rewriting Framework



- A rewrite theory is a triple (Σ, E, R) , with (Σ, E) an equational theory with Σ a signature of operations and sorts, and E a set of (possibly conditional) equations, and with R a set of (possibly conditional) rewrite rules.
- A few logical predicates for our natural languages representations: (sub)concepts, roles, contexts and a few relations between these.
- But the concepts that the representations would use in a minimally working system in the order of 135 thousand, concepts in WordNet. Scaling issues?





Maude Rewriting

- Basic rewriting sorts: Relations, SBasic and UnifSet
- TIL basic assertions such as `canary ⊑ bird` belong to Relations.
- Concept and contextual assertions, such as `instantiable(drink-0,t)` belong to the SBasic basic statements sort.
- The third basic sort, UnifSet, contains unification of skolem constants, such as `crow-6 := bird-1`. This last sort is necessary for unifying skolem constants.



Experimental Results: a few theorems

- 1. a crow was thirsty. ⊢ a thirsty crow
- 2. a thirsty crow ⊢ a crow
- 3. ed arrived and the crow flew away. ⊢ the crow flew away
- 4. ed knew that the crow slept ⊢ the crow slept
- 5. ed did not forget to force the crow to fly ⊢ the crow flew
- 6. the crow came out in search of water ⊢ the crow came out
- 7. a crow was thirsty ⊢ a bird was thirsty

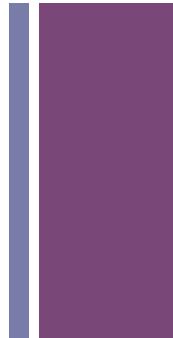


Conclusions

- Proof-of-concept framework
- Introduced a general rewriting framework, using KIML assertions and TIL inference system for textual entailment
- Demonstrated by example that framework can be implemented in Maude and used it to prove in an semi-automated fashion whether a sentence follows from another
- 'shallow theorem proving' for common sense applications?
- Many problems: black box, ambiguity, temporal information, etc..

+

Thanks!



+

Thanks!





References

Revisiting a Brazilian Wordnet. Valeria de Paiva, Alexandre Rademaker, (2012)

Proceedings of Global Wordnet Conference, Global Wordnet Association, Matsue.

OpenWordNet-PT: An Open Brazilian WordNet For Reasoning. de Paiva, Valeria, Alexandre Rademaker, and Gerard de Melo. In *Proceedings of the 24th International Conference On Computational Linguistics*. <http://hdl.handle.net/10438/10274>.

OpenWordNet-PT: A Project Report. Alexandre Rademaker, Valeria de Paiva, Gerard de Melo, Livy Real and Maira Gatti.

Proceedings of the 7th Global Wordnet Conference, Tartu, Estonia. Global Wordnet Association, 2014.

Embedding NomLex-BR Nominalizations Into OpenWordnet-PT. Coelho, Livy Maria Real, Alexandre Rademaker, Valeria De Paiva, and Gerard de Melo. 2014. In *Proceedings of the 7th Global WordNet Conference*. Tartu, Estonia



References

Towards a Universal Wordnet by Learning from Combined Evidence Gerard de Melo, Gerhard Weikum (2009)

18th ACM Conference on Information and Knowledge Management (CIKM 2009), Hong Kong, China.

Bridges from Language to Logic: Concepts, Contexts and Ontologies Valeria de Paiva (2010)

Logical and Semantic Frameworks with Applications, LSFA'10, Natal, Brazil, 2010.

'A Basic Logic for Textual inference", AAAI Workshop on Inference for Textual Question Answering, 2005.

``Textual Inference Logic: Take Two", CONTEXT 2007.

``Precision-focused Textual Inference", Workshop on Textual Entailment and Paraphrasing, 2007.

PARC's Bridge and Question Answering System Proceedings of Grammar Engineering Across Frameworks, 2007.



How do we got about it?

Totally unbaked ideas...

- The future seems easier if it's Open Source (see Ann Copestake's page)
- And collaborative (that too!)
- Translation and comparison of results is necessary
- Many more lexical resources need to be created and shared
- Machine learning of semantics/kr is required
- Logics, building up from ECD, using probabilistic component need to be in place
- Looking on the bright side... LOTS of FUN WORK!



A bridge between language and logic

Wish List:

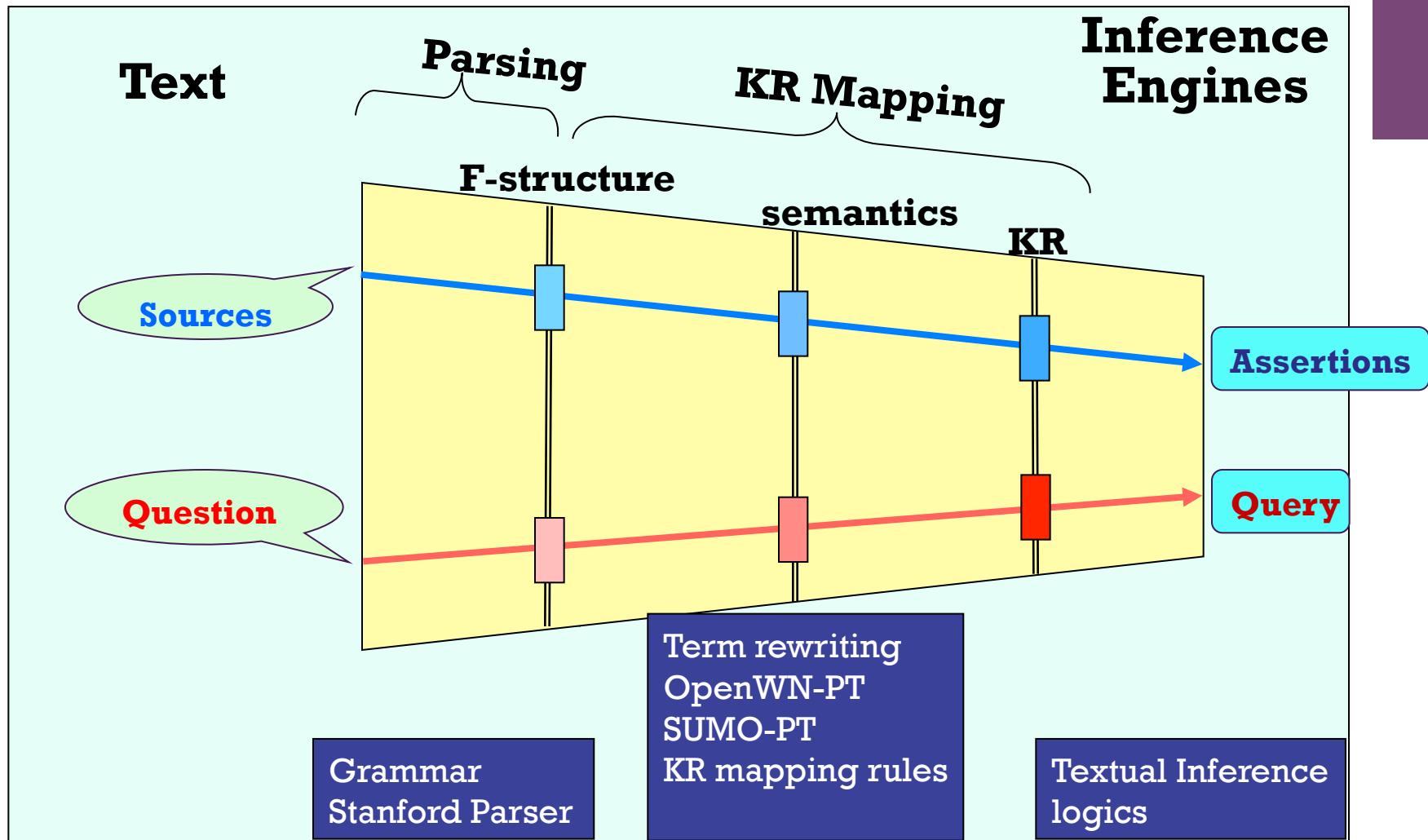
- translation compositional and principled,
- meaning preserving, at least truth value preserving...
- a reasonable fragment of all language
- generic texts
- “logical forms” obtained are useful for reasoning.

Questions:

- which kind of logic on the target?
- how do we know when we’re done?
- how do we measure quality of results?



+ Simplifying the PARC's Bridge Architecture



Idea: Simplify and reproduce components in PORTUGUESE