

TRIOS-TimeBank Corpus: Extended TimeBank corpus with help of Deep Understanding of Text

The seventh international conference on Language
Resources and Evaluation (LREC), Malta, 2010

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TimeML, TimeBank, TempEval-1 and TempEval-2

- TimeML is the scheme for temporal annotation by James Pustejovsky et al.
- TimeBank is the first annotated corpus
- TempEval 1 and 2 are annotated corpus and was shared task on temporal information extraction

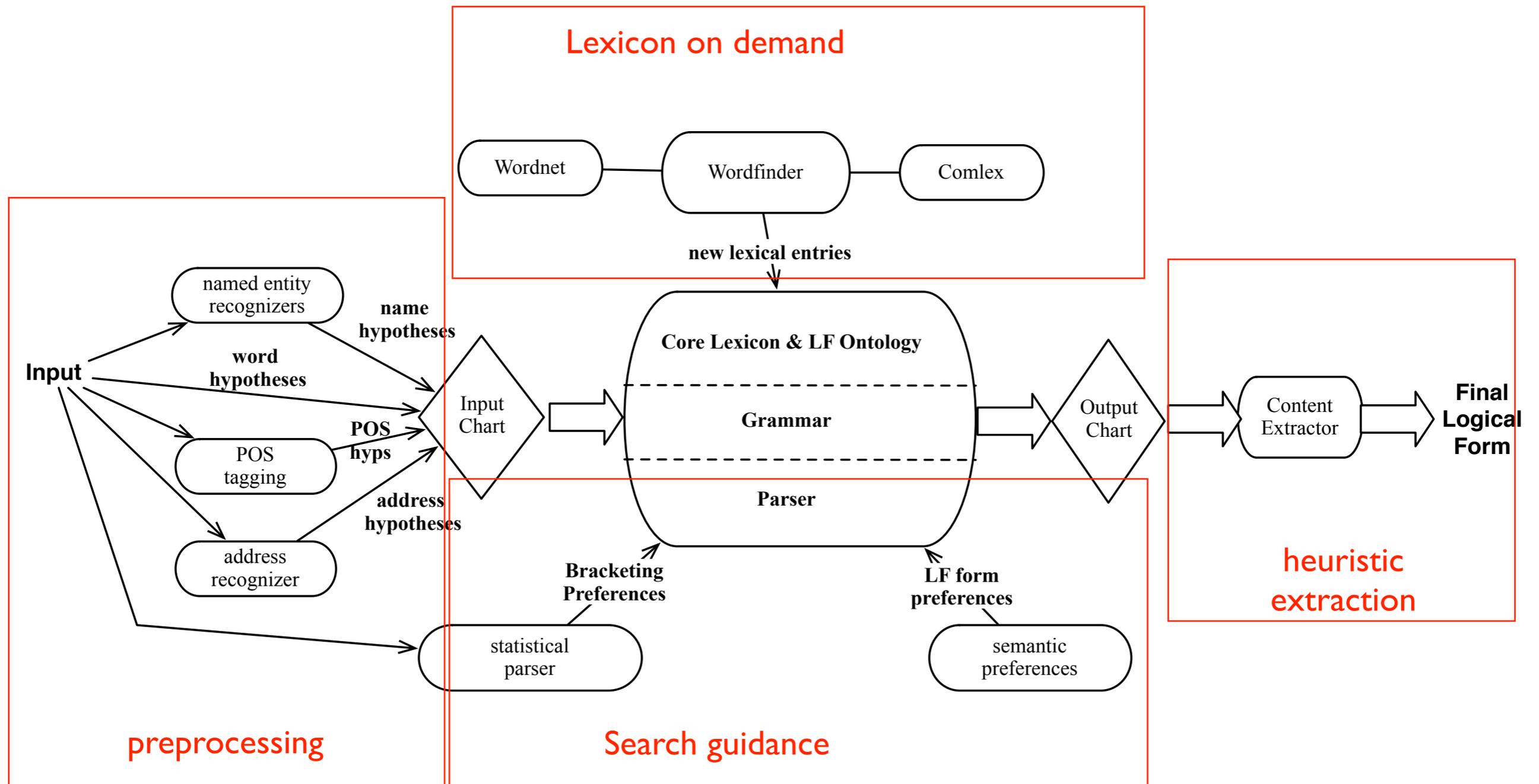


Outline

- **Our System**
- Extension to TimeBank and TimeML
 - Suggest new event
 - Suggest new temporal expressions
 - Suggest ontology type as new event features
 - Suggest improved relations in TimeML
- Future Work



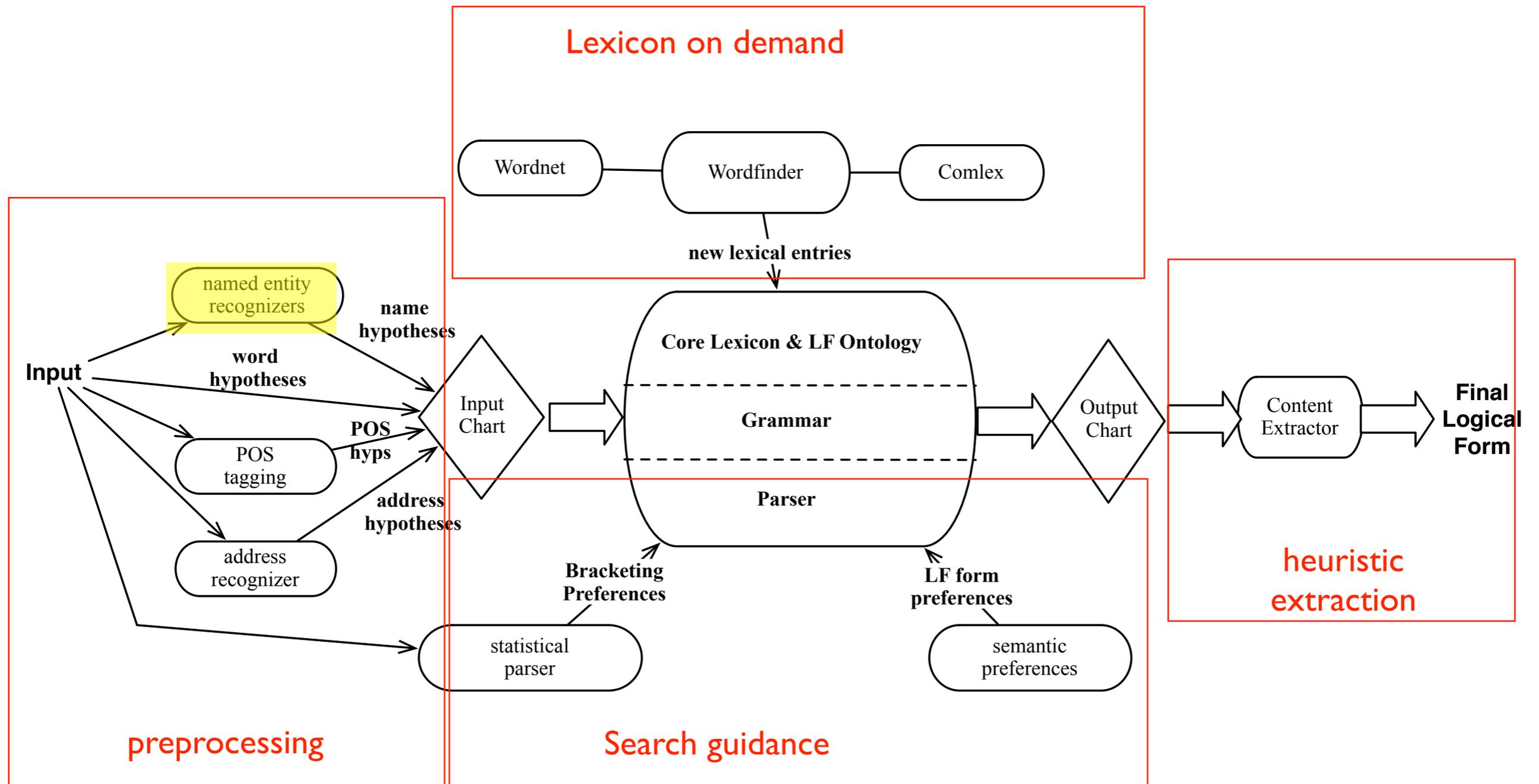
TRIPS Parser: Broad coverage deep parsing



"Advanced Medical paid \$ 106 million in cash for its share in a unit of Henley 's Fisher Scientific subsidiary .



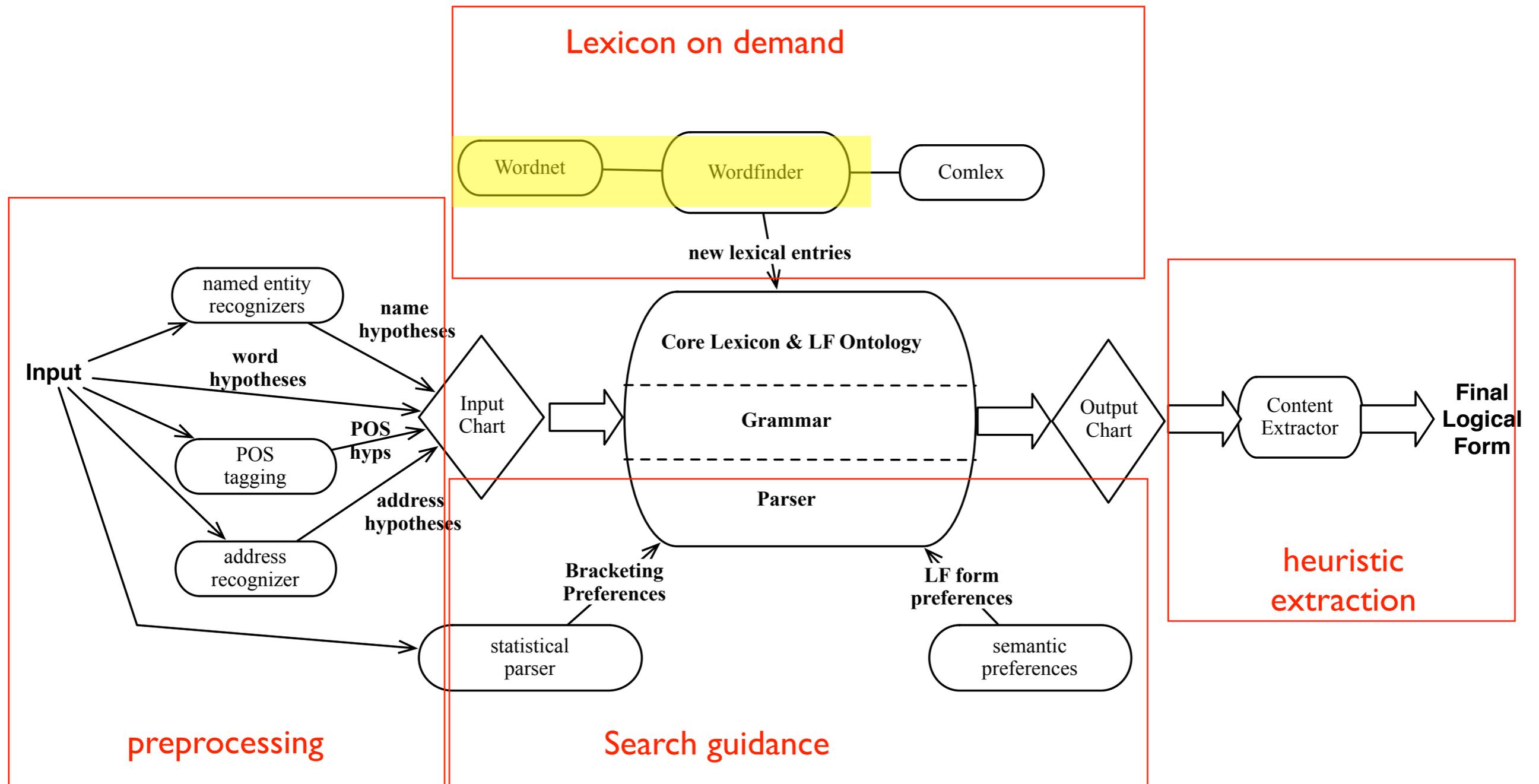
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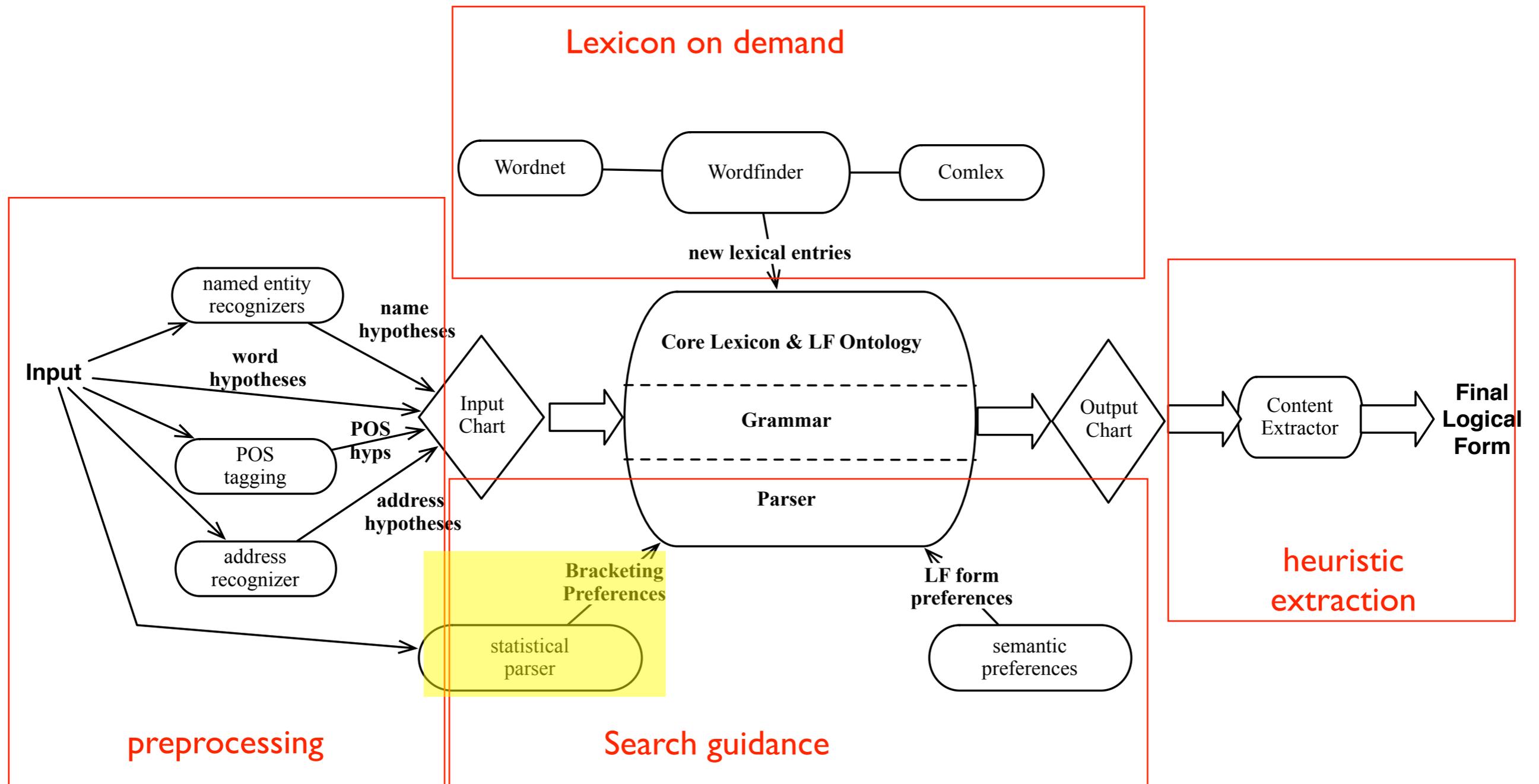
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TRIPS Parser: Broad coverage deep parsing



"Advanced Medical paid \$ 106 million in cash for its share in a unit of Henley 's Fisher Scientific subsidiary .



Events and event features extraction using TRIPS parser

Sentence He fought in the war

TRIPS parser
output

```
(SPEECHACT V1 SA-TELL :CONTENT V2)
(F V2 (:* FIGHTING FIGHT) :AGENT V3 :MODS (V4) :TMA ((TENSE PAST)))
(PRO V3 (:* PERSON HE) :CONTEXT-REL HE)
(F V4 (:* SITUATED-IN IN) :OF V2 :VAL V5) (THE V5 (:* ACTION WAR))
```

100+
Extraction rules

```
((THE ?x (? type SITUATION-ROOT))
  -extract-noms>
  (EVENT ?x (? type SITUATION-ROOT) :pos NOUN :class OCCURRENCE ))
```

Extracted with
extraction rules

```
<EVENT eid=V2 word=FIGHT pos=VERBAL ont-type=FIGHTING tense=PAST
  class=OCCURRENCE voice=ACTIVE aspect=NONE polarity=POSITIVE nf-morph=NONE>
<RLINK eventInstanceID=V2 ref-word=HE ref-ont-type=PERSON relType=AGENT>
<SLINK signal=IN eventInstanceID=V2 subordinatedEventInstance=V5
  relType=SITUATED-IN>
<EVENT eid=V5 word=WAR pos=NOUN ont-type=ACTION class=OCCURRENCE
  voice=ACTIVE polarity=POSITIVE aspect=NONE tense=NONE>
```



Event Extraction Performance

TempEval-1 and
TimeBank

System	Precision	Recall	Fscore	(P+R)/2
TRIOS avg	0.8638	0.7074	0.7778	0.7856
TRIPS avg	0.5801	0.8513	0.6900	0.7157
STEP	0.82	0.706	0.7587	0.763
Sim-Evita	0.812	0.657	0.727	0.7345
IAA	N/A	N/A	N/A	0.78

Table 1: Event Extraction Performance on Bethard and Martin's test data

TempEval-2

System	Precision	Recall	Fscore
TRIOS	0.80	0.74	0.77
TRIPS	0.55	0.88	0.68
Best (TIPSem)	0.81	0.86	0.84

Table 1: Performance of Event Extraction (Task B) in TempEval-2



Performance on Temporal Expression Extraction

		TRIPS TRIOS	Best HeidelTime-1
Temp Exp extraction	Precision	0.85	0.90
	Recall	0.85	0.82
	Fscore	0.85	0.86
Normalization	<i>type</i>	0.94	0.96
	<i>value</i>	0.76	0.85

Table 1: Performance on Temporal Expression extraction (Task A)



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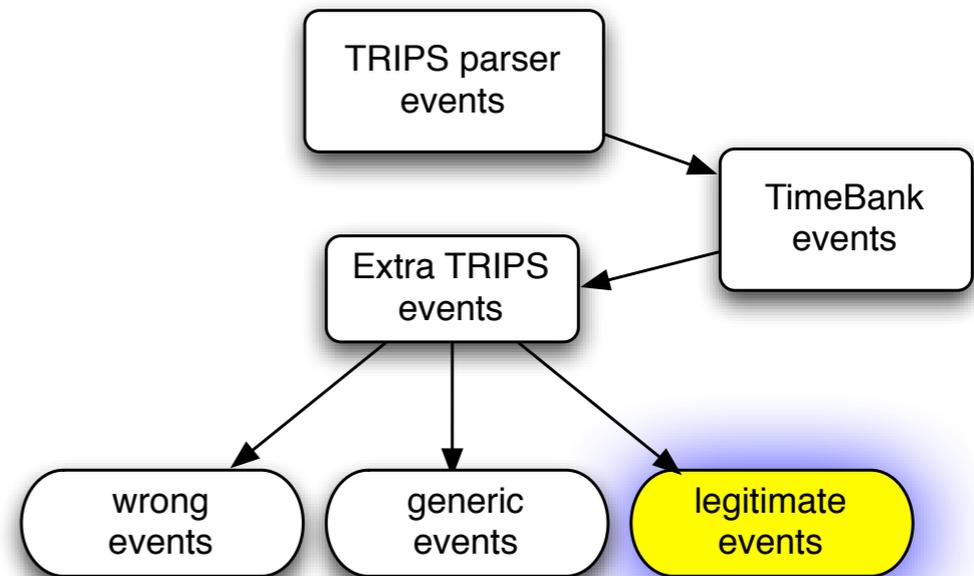
What not to tag?

- Not to tag generic interpretations
 - *Use of corporate jets for political travel is legal.*
- Complements are generics
 - *He said students are prohibited from fighting with each other.*
- Nominalization without extra information
 - *Newspaper reports have **said** ...*



TRIOS minus TimeBank

- Result of wrong parse
- Generic event
- Legitimate event but missed by annotators



Missed legitimate events

1. *At least one of the sensitive sites was a barracks of the elite Republican Guard, a well-placed source **told** The Associated Press.*
2. *Net interest income for the third quarter **declined** to \$35.6 million from \$70.1 million a year ago.*
3. *About \$518 million of debt is **affected**.*
4. *If Iraq **chooses** a simple war of nerves and economic attrition, the Bush administration **knows** a long stalemate could try the patience of the American public and the West in general, and could **open** the possibility that moderate Arabs -- even including Saudi Arabia -- might **drop** out of the effort against Iraq and **accept** some deal from Saddam Hussein.*
5. *"It's the whole uncertainty about what's happening around us," **said** Valentin Von Korff, a trader at Credit Suisse First Boston in Frankfurt.*



Suggest new verbal events to TimeBank

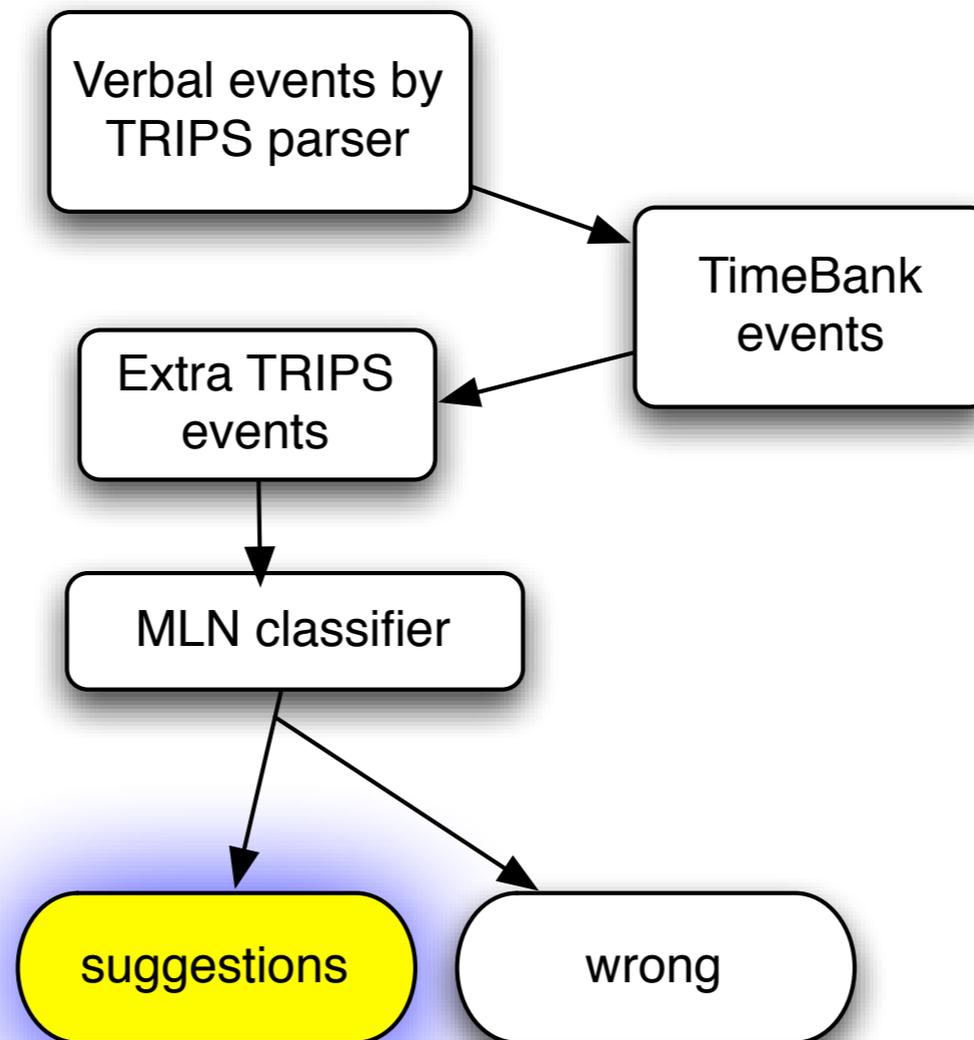


Table 1: Performance on suggested events

TimeML spec	Our Analysis	Number	Performance
Correct	Correct	90	60%
	Generic complement	5	3.3%
Generic	Generic	28	18.8%
	True throughout	8	5.3%
	Others	2	1.3%
Wrong	Wrong	13	8.7%
	Not appropriate	3	2%

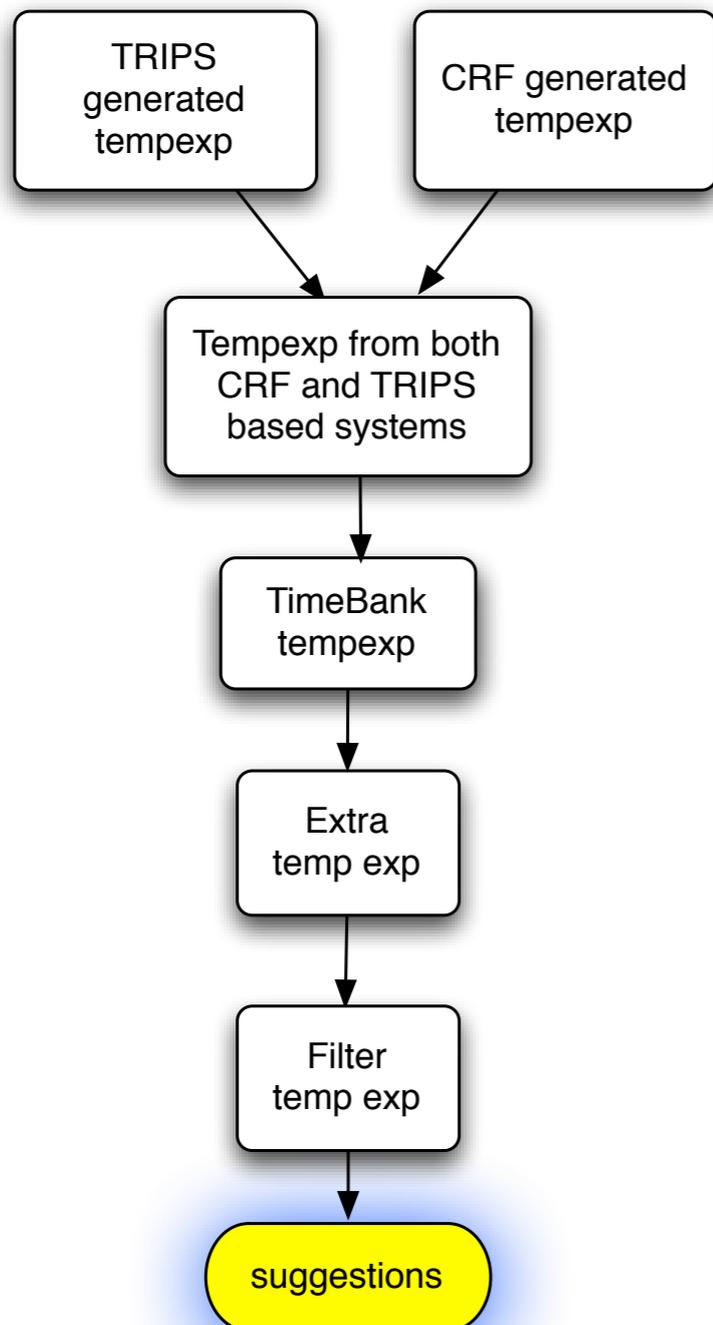
**Total event: 7935
suggestion: ~1%**

true-throughout

If Iraq chooses a simple war of nerves and economic attrition, the Bush administration **knows** a long stalemate could try the patience of the American public and the West in general, ...



Suggest new temporal expressions



Total timex: 1414
accuracy: 50/68 = 73.5%
suggestion: ~3.5%

- *At the end of the broadcast this evening, one more trip around Havana to see what it's been like since **the last time**.*
- *Turks feel they have special ties to the whole region, which they ruled for **hundreds of years** during the Ottoman Empire.*
- *Weisfield's, based in Seattle, Wash., **currently** operates 87 specialty jewelry stores in nine states.*
- ***Previously**, watch imports were denied such duty-free treatment.*



Add ontology-type as new event feature

- TimeML captures event information with coarse-grained *class* (7) or *pos*, or fine-grained *word*
- TRIPS Ontology type
 - more fine-grained than *class* or *pos*
 - coarse-grained than *word*
 - *Fighting* for *fight*, *Action* for *war*
 - Few other words with ont-type *Fighting*: contend, defend, struggle, etc.
- Mapping to WordNet
- Freely available

TimeML Class:

1) *Occurrence*: die, crash, build;

2) *State*: on board, kidnapped;

3) *Reporting*: say, report;

4) *I-Action*: attempt, try, promise;

5) *I-State*: believe, intend, want;

6) *Aspectual*: begin, stop, continue;

7) *Perception*: see, hear, watch, feel.



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More SLINK instances

- SLINK or Subordinate Links: relation between two events
- TimeML includes: modal, factive, counterfactive, evidential, negative evidential and conditional
- Many other cases when one event is argument of other



SLINK examples

- "They have to **continue** to tighten their belts," said Craig Kloner, an analyst at Goldman, Sachs amp Co. (Purpose)
- He **fought** in the war.

```
<SLINK signal=IN eventInstanceID=V2  
subordinatedEventInstance=V5 relType=SITUATED-IN>
```

- Suggest ~900 SLINKs



New Relation Link, RLINK

- Dependency information improves temporal relation identification performance (Chambers et al. 2007 and Katsumasa et al. 2009)

- Chambers and Jurafsky (2008)'s narrative chain with AGENT

```
<RLINK eventInstanceID=V2 ref-word=HE ref-ont-  
type=PERSON relType=AGENT>
```

- Suggest ~2000 RLINKs



Table 1: Most common relTypes used in SLINKs and RLINKs

Our Role	VerbNet equivalents	Lirics equivalents	SLINK Count	RLINK Count
Agent	Agent, Actor	Agent	19	709
Theme	Theme, Stimulus	Theme	336	1137
Affected	Patient	Patient	13	92
Cause	Cause	Cause		49
Goal-as-Loc	Destination	finalLocation	47	
To-Loc	Recipient	Goal	46	
At-Loc	Location	Location	42	
In-Loc	Location	Location	28	
On	Location	Location	20	
Situated-In	Location?	Location?	39	
Purpose	—	Purpose	226	



Approaching TempEval-2

Features	TE2 C	TE2 D	TE2 E	TE2 F
	TE1 A	TE1 B	TE1 C	
<i>Event Class</i>	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$ ¹
<i>Event Tense</i>	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Event Aspect</i>	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Event Polarity</i>	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Event Stem</i>	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Event Word</i>	YES	YES	YES	YES
<i>Event Constituent</i> ²		YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Event Ont-type</i> ³	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Event LexAspect</i> ⁴ \times <i>Tense</i>	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Event Pos</i>	YES	YES	$e_1 \times e_2$	$e_1 \times e_2$
<i>Timex Word</i>		YES		
<i>Timex Type</i>	YES	YES		
<i>Timex Value</i>	YES	YES		
<i>Timex DCT relation</i>	YES	YES		
<i>Event's semantic role</i> ⁵	YES	YES		
<i>Event's argument's ont-type</i>	YES	YES		
<i>TLINK event-time signal</i> ⁶	YES	YES		
<i>SLINK event-event relation type</i> ⁷				$e_1 \times e_2$

Table 1: Features used for TempEval-2 (TE2) Task C, D, E and F or TempEval-1 (TE1) Task A, B and C.



Performance in TempEval-2

Task	TRIPS		TRIOS		Best (with corpus features)
	Precision	Recall	Precision	Recall	Precision
Task C	0.63	0.52	0.65	0.52	0.63 (JU-CSE, UCFD, NCSU-indi)
Task D	0.76	0.69	0.79	0.67	0.82 (TIPSem)
Task E	0.58	0.50	0.56	0.42	0.55 (TIPSem)
Task F	0.59	0.54	0.6	0.46	0.66 (NCSU-individual)

Table 1: Performance of Temporal Relations on TempEval-2 (Task C-F)

Task	Description	Best
Task A	Temporal expression extraction	TRIOS
Task B	Event Extraction	TIPSem
Task C	Event-Timex relationship	TRIOS
Task D	Event-DCT relationship	TIPSem
Task E	Main event-event relationship	TRIOS
Task F	Subordinate event-event relationship	TRIOS

Table 1: Head-to-head comparison of TRIOS, TIPSem and JU-CSE-TEMP (teams that approached all tasks) in TempEval-2 challenge

Naushad UzZaman and James Allen. TRIPS and TRIOS System for TempEval-2: Extracting Temporal Information from Text. SemEval-2 Workshop at ACL-2007.



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Suggest Temporal Links

- Suggest new event-time temporal links that are missed by TimeBank annotators
- Suggest intra-sentence event-event temporal relations, which are ignored in TempEval-1



Automatically Building Larger Temporally Annotated Corpora

- Automatically build larger temporally annotated corpus for news domain, which can be reviewed by human annotators
- Automatically build temporally annotated corpus for other domains like medical domain



Summary

- Suggested New Events in TimeBank
- Added New Event Feature - *Ontology Type*; released TRIPS ontology
- Suggested New Temporal Expressions in TimeBank
- Added Improved Relations in Existing Annotation Scheme
 - SLINK or Subordinate links - relation between events
 - RLINK or Relation link - relation between event and its arguments
- Released TRIOS-TimeBank corpus, TRIPS ontology



Resources

- TRIOS-TimeBank Corpus & TRIPS Ontology:
<http://www.cs.rochester.edu/u/naushad/trios-timebank-corpus>
- Other temporal related resources:
<http://www.cs.rochester.edu/u/naushad/temporal>



Questions ?

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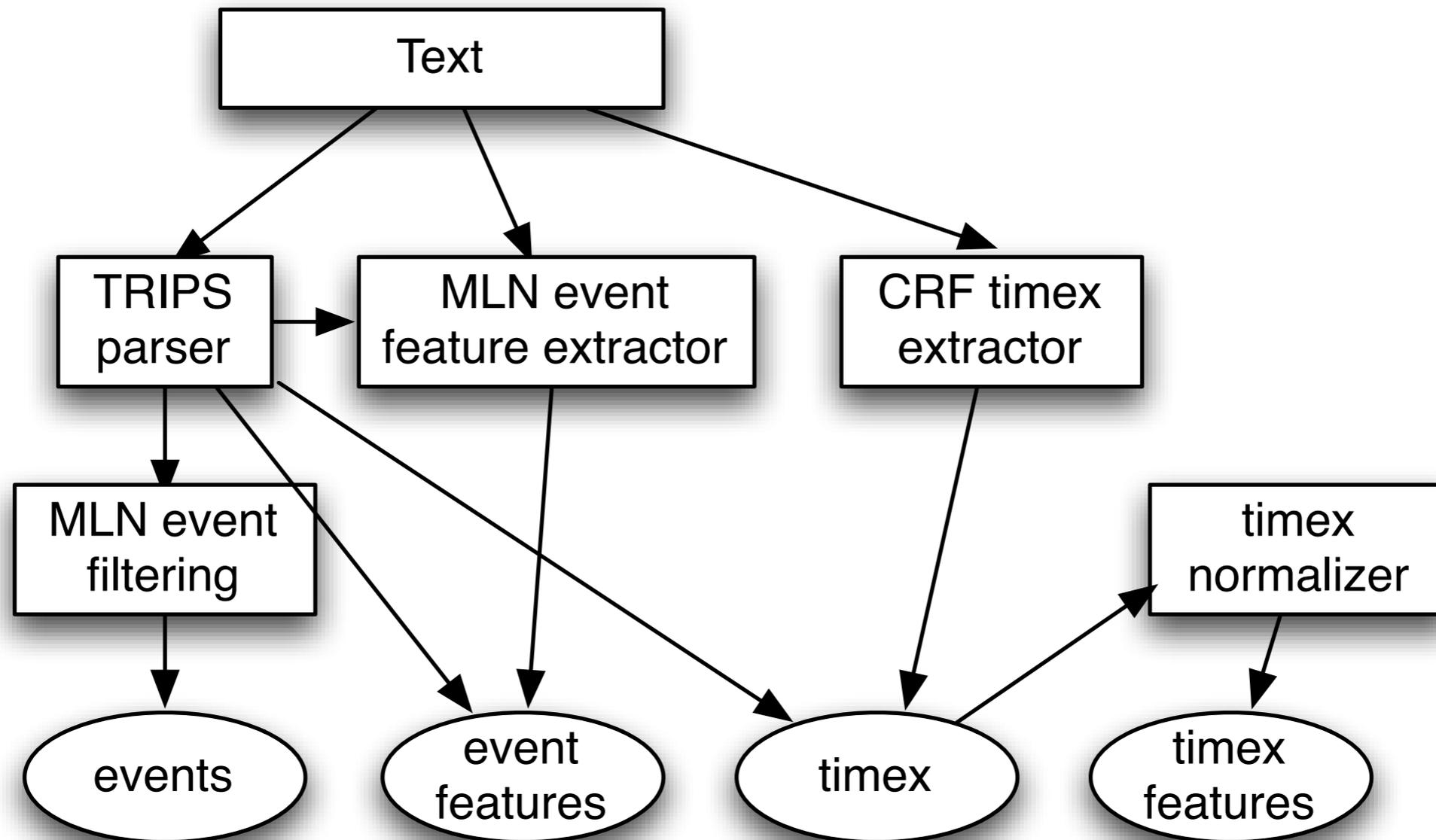
Event Feature Extraction Performance

System	TRIPS	TRIOS	Best
<i>Class</i>	0.67	0.77	0.79 (TIPSem)
<i>Tense</i>	0.67	0.91	0.92 (Edinburgh-LTG)
<i>Aspect</i>	0.97	0.98	0.98
<i>Pos</i>	0.88	0.96	0.97 (TIPSem, Edinburgh-LTG)
<i>Polarity</i>	0.99	0.99	0.99
<i>Modality</i>	0.95	0.95	0.99 (Edinburgh-LTG)

Table 1: Performance of Event Features on TempEval-2 (Task B)



Overview



Benefits of TRIPS ontology

- Superior semantic ontology; better abstraction
- No problem with word sense disambiguation
- Considers semantic roles for disambiguation
- Helps to generate better links



Why RLINK

- RLINKs could be in a separate layer
- needed for complete temporally aware system
- hence included for a complete temporally annotated corpus



Markov Logic Network

- Problems with rule based system and machine learning techniques
- Markov logic = first order logic + markov network (probabilistic graphical model)
 - FOL with weights
 - weights determine how much penalty for a formula to be violated

Example: It is not going to change

`tense(e1, INFINITIVE) & aspect(e1, NONE) => class(e1, OCCURRENCE) weight = 0.319913`