Introduction

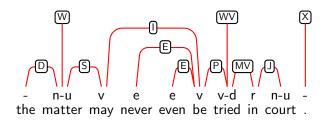
Deriving Multi-Headed Planar Dependency Parses from Link Grammar Parses

Juneki Hong, Jason Eisner

January 28, 2015

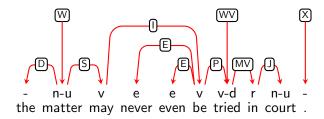
Introduction

- ► This talk is about automatically converting from one annotation style to another.
- ► The conversion could be hard, where information is fragmented or missing.
- ▶ We use a general technique, Integer Linear Programming to recover this missing information.



Link Grammar: Parse with undirected edges

What We Wanted:



Multiheaded parse with directionalized edges

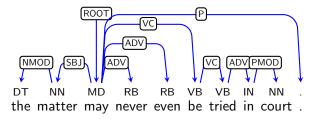
Conclusions

Why We Wanted That

- We want to develop parsing algorithms for parses that look like this
- ▶ We couldn't figure out where to get the data to test them.

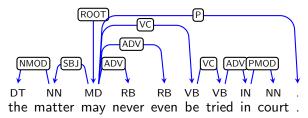
Single-headedness

- Dependency parse treebanks today are either single-headed or not planar.
- Stanford Dependencies are multiheaded but not planar



Some example dependency parse.

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Some example dependency parse.

Link Grammar is almost a multiheaded planar corpora! We just need to directionalize the links.

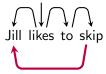
Why Multi-headedness?

Multi-headedness Can Capture Additional Linguistic Phenomenon

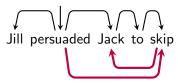
- Control
- Relativization
- ► Conjunction

Control

Control



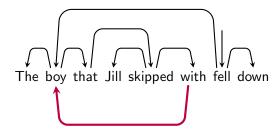
Jill is the subject of two verbs



Jack is the object of one verb and the subject of another

Relativization

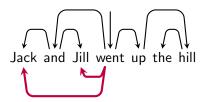
Relativization



The boy is the object of with as well as the subject of fell.

Conjunction

Conjunction



Jack and Jill serve as the two arguments to and, but are also subjects of went.

ntroduction	Motivation	Link Grammars	ILP Model	Experiments and Results	Conclusions
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Motivation

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- We explored whether the Link Grammar could be adapted for this purpose.

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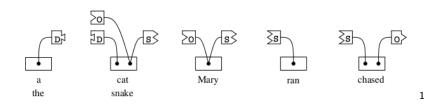
- A multiheaded dependency corpus would be useful for testing new parsing algorithms
- Such a corpus could be automatically annotated using Integer Linear Programming
- We explored whether the Link Grammar could be adapted for this purpose.
- ▶ The results of this are mixed, but provides a good case study.

Link Grammars

Grammar-based formalism for projective dependency parsing with undirected links

Original formalism and English Link Grammar created by Davy Temperley, Daniel Sleator, and John Lafferty (1991)

Link Grammars: How They Work

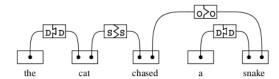


¹These figures were clipped from the original Link Grammar paper:

[&]quot;Parsing English with a Link Grammar" by Sleator and Temperley

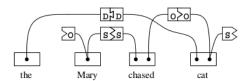
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Link Grammars: How They Work

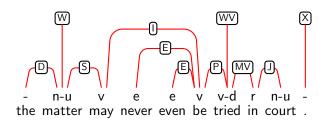


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Link Grammars: How They Work

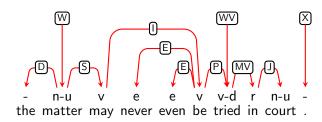


Link Grammars: Same Example Parse From Before Again



Link Parse of a sentence from Penn Tree Bank

Link Grammars: Converting into a Directed Acyclic Graph

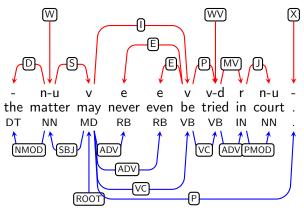


Directionalize the edges

ntroduction	Motivation	Link Grammars	ILP Model	Experiments and Results	Conclusions
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Link Grammars

Compare resulting dependency parse with CoNLL 2007 shared task.

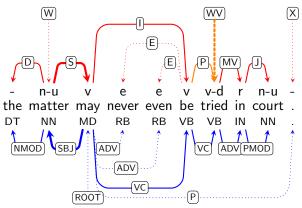


Bottom half is CoNLL. Top half is the directionalized link parse.

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Link Grammars

Compare resulting dependency parse with CoNLL 2007 shared task.



Bottom half is CoNLL. Top half is the directionalized link parse.

What is ILP?

Introduction

What is Integer Linear Programming?

An optimization problem where some or all of the variables are integers.

ILP Model

- The objective function and constraints are linear.
- ▶ In general, it's NP-Hard! But good solvers exist that work well most of the time.
- ▶ Our ILP is encoded as a ZIMPL program and solved using the SCIP Optimization Suite²

²http://scip.zib.de/

Corpus Building Strategy

- We start with a corpus of link grammar parses.
- We try to directionalize this corpus, choosing a direction for every link in every parse.
- We use an Integer Linear Program to find consistent directions for every link

Integer Linear Programming Model

Encoded Constraints:

- Acyclicity
- Connectedness
- Consistency of Directionalized Links

Integer Linear Programming Model

Encoded Constraints:

- Acyclicity: No cycles!
- ► Connectedness: Every word is reachable from a root
- Consistency of Directionalized Links: Similar links oriented the same way

Integer Linear Programming Model

For each sentence, for each edge i, j, where i < j



Variables:

$$x_{ij}, x_{ji} \in \mathbb{Z} \geq 0$$
: orientation of each link $x_{ij} + x_{ji} = 1$

Integer Linear Programming Model

For each sentence, for each edge i, j, where i < j



Variables:

$$x_{ij}, x_{ji} \in \mathbb{Z} \geq 0$$
: orientation of each link $x_{ii} + x_{ji} = 1$

A link can either be oriented left or oriented right

Acyclicity, Connectedness

Acyclicity

Given that node u is the parent of v

 n_v : length of the sentence containing node v

 $d_{\nu} \in [0, n_{\nu}]$: depth of the node from the root of the sentence

$$(\forall_u) \ d_v + (1 + n_v) \cdot (1 - x_{uv}) \ge 1 + d_u$$
 (1)

Connectedness

$$\sum x_{uv} \ge 1 \tag{2}$$

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Acyclicity

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$$(\forall_u) d_v + (1 + n_v) \cdot (1 - x_{uv}) \ge 1 + d_u$$
 (1)

The depth of a child is greater than the depth of the parent Connectedness

$$\sum x_{uv} \ge 1 \tag{2}$$

A word has at least 1 parent

Consistency of Directionalized Links

Consistency of Directionalized Links

 $r_L, \ell_L \in \{0, 1\}$: whether links with label L allowed left/right

$$x_{ij} \le r_L$$
 $x_{ji} \le \ell_L$ (3)

$$\min\left(\sum_{L}r_{L}+\ell_{L}\right) \tag{4}$$



ILP Model ○ ○ ○ Experiments and Results

Conclusions

ILP Model

Introduction

Consistency of Directionalized Links with Slack

Consistency of Directionalized Links

 $r_{l}, \ell_{l} \in \{0,1\}$: whether links with label L allowed left/right

$$x_{ij} \leq r_L + s_{ij} \qquad \qquad x_{ji} \leq \ell_L + s_{ij} \tag{3}$$

$$\min\left(\sum_{L}r_{L}+\ell_{L}\right)\cdot\frac{N_{L}}{4}+\sum_{ij}s_{ij}\tag{4}$$

 $s_{ii} \in \mathbb{R} \geq 0$: slack variable

 N_I : Number of link tokens with label L

Slack allows a few links with label L in disallowed directions

Data Sets

Data Sets taken from:

CoNLL 2007 Shared Task (English)

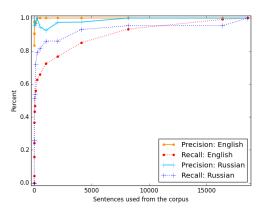
ACL 2013 Shared Task of Machine Translation (Russian)

	Input Sentences	Output Connected Parses
English	18,577	10,960
Russian	18,577	4,913

Stability of Results

Introduction

- We worried that the recovered direction mapping might be unstable and sensitive to the input corpus.
- We compared the results of increasing runs of sentences.



On the English Data Set:

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Multiheadedness

Link Data has 8% additional edges over the CoNLL. (average about 2 multiheaded words per sentence)

Conclusions

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Multiheadedness

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CoNLL Matches

52% of links match CoNLL arcs

57% of CoNLL arcs match links

Multiheadedness

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Directionality

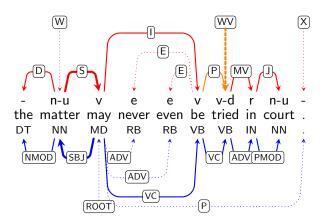
6.19% of link types allowed both directions

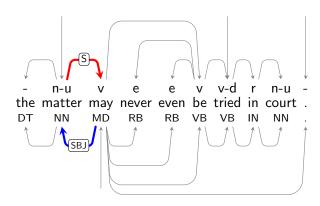
2.07% of link tokens required disallowed direction via slack

Introduction

Label	Rightward	Multiheaded	CoNLL Match	CoNLL Dir Match
Α	0% (0/8501)	0% (0/8501)	84% (7148/8501)	98% (7002/7148)
AN	0% (0/9401)	0% (0/9401)	83% (7825/9401)	98% (7639/7825)
В	100% (1514/1515)	61% (919/1515)	53% (806/1515)	84% (678/806)
C	100% (3272/3272)	0% (0/3272)	3% (85/3272)	53% (45/85)
CO	0% (0/2478)	1% (32/2478)	5% (114/2478)	68% (78/114)
CV	100% (3237/3237)	100% (3237/3237)	56% (1827/3237)	28% (512/1827)
D	0% (56/19535)	0% (71/19535)	85% (16656/19535)	100% (16608/16656)
E	0% (0/1897)	0% (2/1897)	67% (1279/1897)	99% (1263/1279)
G	0% (0/6061)	0% (0/6061)	70% (4258/6061)	96% (4070/4258)
1	100% (5405/5424)	60% (3247/5424)	95% (5168/5424)	47% (2408/5168)
IV	100% (1626/1627)	100% (1626/1627)	85% (1389/1627)	97% (1353/1389)
J	98% (16400/16673)	2% (280/16673)	87% (14522/16673)	97% (14069/14522)
M	100% (9594/9596)	0% (16/9596)	74% (7124/9596)	92% (6583/7124)
MV	100% (13375/13376)	0% (61/13376)	51% (6797/13376)	98% (6681/6797)
MX	100% (1999/1999)	4% (83/1999)	42% (836/1999)	91% (763/836)
0	100% (11027/11028)	0% (0/11028)	81% (8932/11028)	96% (8535/8932)
P	100% (3755/3756)	31% (1167/3756)	94% (3528/3756)	100% (3523/3528)
S	97% (13138/13520)	57% (7662/13520)	92% (12476/13520)	5% (586/12476)
SJ	50% (2736/5468)	0% (0/5468)	69% (3778/5468)	93% (3502/3778)
TO	100% (1733/1734)	0% (1/1734)	0% (5/1734)	100% (5/5)
VJ	51% (765/1500)	1% (8/1500)	71% (1059/1500)	89% (939/1059)
W	100% (10528/10528)	0% (5/10528)	5% (504/10528)	46% (232/504)
WV	100% (7563/7563)	100% (7557/7563)	57% (4345/7563)	97% (4214/4345)
X	80% (13132/16406)	5% (806/16406)	8% (1364/16406)	95% (1300/1364)
YS	0% (0/1645)	0% (0/1645)	98% (1619/1645)	0% (0/1619)

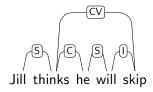
ILP Model

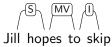




Link Results: Subject-Verb Links are Backwards

► This is due to a possible inconsistency of the Link Grammar, discovered by our method.





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- But the governing verb usually links to the subject when there is one.
- ► So this makes the subject a consistent choice to make the head of a clause.

To fix this, we could edit the link grammar, link parses, or the ILP.

Conclusions

- ▶ Link Grammar parses can be oriented into connected DAGs
- new corpora for building multi-headed dependency parsers
- ▶ ILP can be used to help annotate missing data in corpora.

Introduction	Motivation	Link Grammars	ILP Model	Experiments and Results	Conclusions
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Questions?