

Coreference relations in EN, DE and RU, and mapping between them

Yulia Grishina

Applied Computational Linguistics FSP Cognitive Sciences University of Potsdam / Germany

Coreference resolution

Daisy Hamilton was a private detective. She was thirty years old and she has been a detective for the past two years. Every morning Daisy went to her office to wait for phone calls or open the door to clients needing her services. One day somebody knocked on the door.

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> Daisy Hamilton the door

language	POS	syntax	coreference
Basque	\checkmark	\checkmark	-
Bulgarian	\checkmark	\checkmark	\checkmark
Catalan	\checkmark	\checkmark	\checkmark
Croatian	\checkmark	\checkmark	-
Czech	\checkmark	\checkmark	-
Danish	\checkmark	\checkmark	-
Dutch	\checkmark	\checkmark	\checkmark
English	\checkmark	\checkmark	\checkmark
Estonian	\checkmark	\checkmark	-
Finnish	\checkmark	\checkmark	-
French	\checkmark	\checkmark	\checkmark
Galician	\checkmark	\checkmark	-
German	\checkmark	\checkmark	\checkmark
Greek	\checkmark	\checkmark	-
Hungarian	\checkmark	\checkmark	\checkmark
Icelandic	\checkmark	\checkmark	-
Irish	\checkmark	\checkmark	-
Italian	\checkmark	\checkmark	\checkmark
Latvian	\checkmark	\checkmark	-
Lithuanian	\checkmark	-	-
Maltese	\checkmark	-	-
Norwegian	\checkmark	\checkmark	\checkmark
Polish	\checkmark	\checkmark	\checkmark
Portuguese	\checkmark	\checkmark	\checkmark

Un^{iversit}äx

Motivation

(Zhekova, 2013)

Idea



High-quality coreference resolvers exist for a small number of languages. How about other languages (in particular, low-resource)?

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-> require multilingual resources

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-> require multilingual resources

-> resource transfer (annotation projection)

Universitär

Outline

(I) direct coreference in English, German, Russian

(II) annotation projection

(III) indirect coreference (bridging & nearidentity)



direct coreference in EN, DE, RU



- 38 parallel texts
- 3 languages: English, German, Russian
- 3 text genres: newswire¹, narratives², medicine instruction leaflets³ (only EN-DE)

¹ multilingual newswire agency Project Syndicate (www.project-syndicate.org)

² short narratives for second language acquisition Daisy stories (http://www.lonweb.org)

³ EMEA subcorpus of the OPUS collection of parallel corpora (Tiedemann, 2009)



Annotation

- common coreference annotation guidelines
- uniform annotations in 3 languages
- related annotation schemes: OntoNotes (Hovy et al., 2006) PoCoS (Krasavina & Chiarcos, 2007), ParCor (Guillou et al., 2014)
- identity relation
- annotation tool: MMAX-2 (Müller & Strube, 2006), subsequently converted into CoNLL-2012 format

Annotation guidelines

- NP coreference: full NPs, proper names, pronouns
- no generic NPs annotated
- no singletons annotated
- \cdot set of attributes defined



Annotation workflow

MMAX2 1.13.003 /Users/Yulia/Documents/CL/Projects/corpus-main/1alesinagiavazzen.mmax

File Settings Display Tools Plugins Info 🗹 Show ML Panel

Europe's] Divided Racial House .

A common feature of [Europe's] extreme right is [[its] racism] and use of the immigration issue as a political wedge .

The Lega Nord in Italy, the Vlaams Blok in the Netherlands, the supporters of [Le Pen's] National Front in France, are all examples of parties or movements formed on the common theme of aversion to immigrants and promotion of simplistic policies to control them.

While individuals like Jorg Haidar and [Jean-Marie Le Pen] may come and (never to soon) go , [the race question] will not disappear from European politics anytime soon .

An aging population at home and ever more open borders imply increasing racial fragmentation in European countries .

[Mainstream parties of the center left and center right] have confronted this prospect by hiding [their] heads in the ground , hoping against hope that [the problem] will disappear .



Annotation workflow

MMAX2	1.13.003	/Users/\	Yulia/Docui	ments/(CL/Projects/corpus-main/1alesinagiavazzen.mmax [modified]
File Settings [Display	Tools	Plugins	Info	✓ Show ML Panel
Europe's] Divided A common feature wedge . The Lega Nord in J	Racial H e of [<mark>Eur</mark> e	ouse . <mark>Spe's] ex</mark> Vlaams	(treme rig Rlok in th	nt is l	[its] racism] and use of the immigration issue as a political retunds the supporters of [Le Pen's] National Front in France [Europe's]
One-click annotation	Panel	Settings			
			primmark	sec	mark groups sentence
referentiality	O no	ot_specifie	d Odisc	:ourse_c	cataphor referring odiscourse-new other
dir_speech	🗿 te	xt_level	direct_s	peech	<pre>indirect_speech</pre>
phrase_type	🗿 nj	o 🔵 pp	other		
np_form	O no	one 🔾	ne 🔵 def	fnp	pper ppos padv pds other
ambiguity	o no	ot_ambig	🔵 ambig	_ante	ambig_rel ambig_rel_ante ambig_idiom ambig_expl
Anaphor_anteceder	nt empty				
< > Type	non	e	0		
complex_np	O no	ot_specifie	d 🔵 yes	() n	0
grammatical_role		ot_specifie	d 🔵 sbj	🔵 di	r-obj 🗿 indir-obj 🗌 other
comment					
				V :	Suppress check 🗸 Warn on extra attributes
			A	pply	Undo changes
				Auto	-apply is OFF



	Newswire			Narratives			Medicine		Total		
	EN	DE	RU	EN	DE	RU	EN	DE	EN	DE	RU
Tokens	5903	6268	5763	2619	2642	2343	3386	3002	11908	11912	8106
Sentences	239	252	239	190	186	192	160	160	589	598	431
REs	558	589	606	470	497	479	322	309	1350	1395	1085
Chains	124	140	140	45	45	48	90	88	259	273	188
Tokens/REs (%)									8.82	8.5	7.47



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Example: news

 NEW YORK – The terrorist sanctuary in [the South Waziristan region] of [Pakistan's] tribal frontier with [Afghanistan] is coming apart. It took a while for [the Pakistani Army] to move against the [region's] rising violence and chaos, but [[its] campaign in [South Waziristan]] is making progress.

-> tokens/RE = 10.58

-> **REs/chain** = 4.5



Example: medicine

- [Abilify] is a medicine containing [the active substance aripiprazole]. [It] is available as 5 mg, 10 mg, 15 mg and 30 mg tablets, as 10 mg, 15 mg and 30 mg orodispersible tablets. [Abilify] is used to treat adults with the following mental illnesses: [schizophrenia, a mental illness with a number of symptoms], ...
 - -> tokens/RE = 10.5
 - -> **REs/chain** = 3.5



Example: narrative

- [Daisy] had got up early that spring morning because [she] was working on a case in the nearby town. [She] arrived at [[her] office] with a paper bag in [her] hand containing fresh cream buns at a quarter to eight and was dying for a cup of coffee.
 - -> tokens/RE = 5.57
 - -> REs/chain = 10.4





- Average number of tokens per RE per lang: EN (8.82), DE (8.5), RU (7.47)
- Average number of tokens per RE: newswire (10.58) > medicine (10.5) > narrative (5.57)
- Average number of REs per chain: narrative (10.4) > newswire (4.5) > medicine (3.5)
- tokens/RE, REs/chain are stable across languages





Inter-annotator agreement

 EN-DE: each text annotated by 2 lightly trained annotators (students of linguistics)

	English	German
Binary overlap κ	0.87	0.86
Proportional overlap κ	0.81	0.81
MUC F-score	77.28	73.91

• EN-RU: only 1 annotator available

Annotation discrepancies

		MMAX2 1.1	3.003 /Us	sers/Yulia/[Docume	ents/CL/Projects/corpus-main/1alesinagiavazzen.mmax
File	Settings	Display	Tools	Plugins	Info	Show ML Panel
Euror A con wedge The L are all promo While will no An ag count [Main the gr	e'sl Divide nmon featu ega Nord i examples otion of sin individual ot disappea ing popula ries . stream pa ound , hop	ed Racial H ure of [Euro n Italy , the of parties nplistic pol s like Jorg ar from Euro ation at ho nties of the bing agains	ouse . ope's] ex e Vlaams or move licies to Haidar a ropean p me and center l st hope t	ctreme rig Blok in the ments for control the nd [Jean- colitics and ever more eft and co	n is ne Net rmed o em . Marie ytime e open	[its] racism] and use of the immigration issue as a political meriands, the supporters of [Le Pen's] National Front in France, on the common theme of aversion to immigrants and Le Pen] may come and (never to soon) go , [the race question] soon . borders imply increasing racial fragmentation in European ant] have confronted this prospect by hiding [their] heads in n] will disappear .

-> need to incorporate near-identity and bridging (part III of the talk)



-> Outcomes I

- created common annotation guidelines
- built a parallel coreference corpus of 3 genres
- compared annotations in 3 languages



II. annotation projection

 automatically transfer annotations from source to target



 automatically transfer annotations from source to target



 automatically transfer annotations from source to target



- POS tags (Yarowsky et al., 2001), (Agic et al., 2015)
- Named Entities (Ehrmann et al., 2011)
- syntactic trees (Tiedemann, 2014), (Johannsen, 2016)
- coreference



Experimental setup

1. automatic sentence and word alignment

2. extraction of REs

3. transfer of coreference chains



Alignment

- Sentence alignment: HunAlign (Varga et al., 2007) and its wrapper LF Aligner
- Word alignment: GIZA++ (Och & Ney, 2003)
 - $\cdot\,$ training set: around 200 000 parallel sentences
 - intersective alignments



Projection

Direct projection algorithm (following (Postolache et al., 2006)):

•

- $\cdot \;\;$ for each word span $e_1 ... e_i$ we extract corresponding aligned words $f_1 ... f_j$
- \cdot remove duplicates, reorder according to the surface order
- target RE is the span between the 1st and the last word and it belongs to the same coreference set as the source RE



Projection

It was a fat lady who wore a fur around her neck. She said that she needs Daisy's help and does not know what to do.

Es war eine dicke Dame mit einer Pelzstola. Sie hat gesagt, dass sie Daisys Hilfe braucht und dass sie nicht weiß, was sie tun soll.


Projection

It was [a fat lady] [who] wore a fur around [her] neck. [She] said that [she] needs [Daisy's] help and does not know what to do.

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It was **[a fat lady] [who]** wore a fur around **[her]** neck. **[She]** said that **[she]** needs **[Daisy's]** help and does not know what to do.

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It was **[a fat lady] [who]** wore a fur around **[her]** neck. **[She]** said that **[she]** needs **[Daisy's]** help and does not know what to do.

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It was **[a fat lady] [who]** wore a fur around **[her]** neck. **[She]** said that **[she]** needs **[Daisy's]** help and does not know what to do.

Es war **[eine dicke Dam** mit einer Pelzstola **[Sie]**

hat gesagt, dass [sie] [Daisys] Hilfe braucht und

dast sie nicht weiß, wat sie tun soll.







	Mentions		MUC		CEAF		B ³			Average (coref)					
	Р	R	F1	Р	R	F1	Р	R	F 1	Р	R	F1	Р	R	F1
de-News	61.5	48.6	54.3	55.9	43.2	48.7	58.6	46.7	51.9	45.8	34.2	39.1	53.4	41.4	46.6
de-Stories	82.0	54.5	65.5	81.9	51.6	63.3	81.7	53. 7	64.8	71.6	32.5	44.7	78.4	45.9	57.6
de-Medicine	61.2	44.7	51.7	66.2	42.7	51.9	59.1	43.3	50.0	53.43	35.16	42.41	59.6	40.4	48.1
de-News _{min}	89.9	71.2	79.4	87.3	66.2	75.3	85.5	67.5	75.5	80.4	58.1	67.5	84.4	63.9	72.8
de-Stories _{min}	95.4	62.2	75.3	94.4	58.5	72.2	95.1	61.2	74.5	90.9	40.2	55.7	93.5	53.3	67.5
de-Medicinemin	79.9	58.4	67.5	84.2	54.4	66.1	77.7	56.9	65.7	73.3	47.2	57.4	78.4	52.8	63.1
ru-News	79.3	64.5	71.2	76.3	60.7	67.6	76.3	62.0	68.4	69.0	52.2	59.4	73.9	58.3	65.1
<i>ru</i> -Stories	87.4	65.1	74.6	87.9	64.4	74.3	86.1	64.6	73.8	79.7	47.9	59.8	84.6	59.0	69.3
ru-News _{min}	90.9	72.6	80.7	89.6	69.8	78.5	87.3	69.7	77.5	83.7	61.4	70.9	86.9	67.0	75.6
ru-Stories _{min}	94.3	72.4	81.9	94.0	70.9	80.9	93.6	71.7	81.2	90.2	57.3	70.1	92.6	66.6	77.4



Results for Russian are better

	Mentions			MUC			CEAF			B^3			Average (coref)		
	Р	R	F 1	Р	R	F 1	Р	R	F1	Р	R	F1	Р	R	F1
de-News	61.5	48.6	54.3	55.9	43.2	48.7	58.6	46.7	51.9	45.8	34.2	39.1	53.4	41.4	46.6
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Results for stories are better



	approach	F1
Postolache et al.	POS tagging, syntactic heads, language-dependent word aligner	63.9
EN-DE	atatiatiaal ward alignmant	54.6
EN-RU	statistical word alignment	71.0



F1-scores for different genres



Error analysis: two-fold



Error analysis: two-fold



Word alignment quality

	Bisents	Ρ	R	F1
Padó (2007)	1,029,400	98.6	52.9	68.86
Spreyer (2011)	1,314,944	94.88	62.04	75.02
Our approach	205,208	92.95	51.23	66.05

Error analysis: two-fold





Typology of errors

(1) Morphological differences

- (1) contractions
- (2) compounds

(2) Differences in NP syntax

- (1) the use of articles
- (2) pre- & post-modification

(3) Non-equivalences in translation

- (1) personal & indefinite pronouns
- (2) relative clauses & participial constructions



(1.1) Compounds

- Europeans, however, are prompt to criticise the US for any failure in their policy toward [minorities].
- Europäer sind allerdings schnell bereit, der USA jeden Fehler in ihrer Minderheitenpolitik vorzuwerfen.

(2.1) The use of articles

- English allows the use of nouns with zero articles more frequently than German (Kunz, 2010)
 - A. Lastly, the G-20 could also help drive momentum on climate change.
 - B. Schließlich könnten die G-20 auch für neue Impulse im Bereich [des Klimawandels] sorgen.
- Named Entities

```
C. Hamas - [die Hamas]
```

(2.2) Pre- & post-modification

- German NPs allow more complicated pre-modification
- A. Pakistan needs international help to bring hope to [the young people] [who] live there.
- B. Pakistan braucht internationale Hilfe, um [den dort lebenden jungen Menschen] Hoffnung zu bringen.
- С. Пакистан нуждается в международной помощи, чтобы дать надежду [молодым людям], [которые] там живут.

(3.1) Personal & indefinite pronouns



- German indefinite pronoun *man*
- A. [It] was pursing a two-pronged strategy.
- B. Man verfolgte eine Doppelstrategie.
- С. [Она] преследовала двойную стратегию.



-> Outcomes II

- \cdot applied annotation projection to 2 language pairs
- knowledge-lean approach
- F1 54.6 (EN-DE), 71.0 (EN-RU)
- \cdot $\,$ errors in alignments and lang divergencies $\,$



|||.

indirect coreference in EN, DE, RU (bridging & near-identity)



Goals

- introduce a typology for bridging relations
- use an existing one for near-identity & apply it to German
- validate on a corpus of different languages and genres



Experiments

- Design for German Apply on German Transfer to English and Russian
- manual transfer

•

 \cdot aiming at automatic projection via parallel corpora

Bridging & near-identity

- Bridging: indirect relations that can only be inferred based on the common knowledge shared by the speaker and the listener (e.g. part-whole, set-membership)
- Near-identity: two NPs are almost identical, but differ in one crucial dimension (e.g. time)



Example: bridging

Daisy walked into **[the office]** and saw a

bunch of flowers on [the windowsill].



Example: bridging

Daisy walked into **[the office]** and saw a

bunch of flowers on [the windowsill].



Example: bridging

*Daisy walked into **[the office]** and saw a

bunch of flowers on the road.

??

Example: near-identity

In the afternoon, **[the temperature]** rose to 20C. This morning **[it]** was 12C.

Example: near-identity

In the afternoon, [the temperature] rose

to 20C. This morning **[it]** was 12C.

-> same referent [temperature], different numerical values

Bridging: 2 viewpoints



Information Status

→ an IS subcategory, along with given, new, etc.(Gardent et al., 2003), (Nissim et al, 2004), (Ritz et al., 2008), (Riester et al., 2010), (Markert et al., 2012)

→ <u>Coreference</u>

→ a separate coreference relation, e.g. part-whole, setmembership (Poesio et al., 2004), (Poesio and Artstein, 2008), (Nedoluzhko et al., 2009)



Annotation

- bridging (Clark, 1975) and near-identity (Recasens et al., 2010)
- German side of the corpus
- \cdot 2 annotators (half of the corpus)
- bridging: examine all definite NPs that are not linked to anything
- near-identity: check all NPs



Parallel corpus

	#EN	#DE	#RU
Documents	14	14	10
Sentences	589	598	431
Tokens	11908	11894	8106
REs	1350	1395	1085
Coreference chains	259	273	188
Bridging markables	188	432	188



Example annotation

[Daisy Hamilton] was a private detective. [She] was thirty years old and [she] has been a detective for the past two years. Every morning [Daisy] went to [[her] office]_B1 to wait for phone calls or open [[the door]_B1] to clients needing [her] services. One day somebody knocked on [the door].

Annotation: bridging



• PART-WHOLE

 \cdot the telephone - the receiver

SET-MEMBERSHIP

• the European Union - the least developed countries

• ENTITY-ATTRIBUTE/FUNCTION

• Kosovo - the current policy of rejection

• EVENT-ATTRIBUTE

 \cdot the regional conflict - the trained fighters

· LOCATION-ATTRIBUTE

• Germany - in the south



Annotation: bridging

- \cdot Annotation principles
 - → semantic relatedness
 - → proximity
 - ➡ identity < near-identity < bridging</p>

[The telephone] rang. I went into [the office] and picked up [the receiver].



Annotation: bridging

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[The telephone] rang.] went into [the office] and picked up [the receiver].


Annotation: bridging

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 - ➡ semantic relatedness
 - → proximity
 - → identity < near-identity < bridging</p>

[The telephone] rang.] went into [the office] and picked up [the receiver].



Results: bridging

	Poesio (2004)	Nedoluzhko et al. (2009)	This work
Anaphor selection (F-1)	0.22	0.5	0.64
Antecedent selection (F-1)	N/A	N/A	0.79
Relation assignment (Cohen's kappa)	N/A	0.9	0.98

Annotation: near-identity

- NAME METONYMY
 - the US (geographical entity) the US (the government)
- MERONYMY
 - the president the US (=the president)
- SPATIO-TEMPORAL FUNCTION
 - Budapest the medieval Budapest

(from Recasens et al., 2010)

Results: near-identity

- small amount of near-identity links in the corpus, insufficient to compute the IAA
- for German, it conforms to the results of (Recasens et al., 2012)
- \cdot -> it is difficult to annotate near-identity explicitly



Relation	News	Narrative	Medicine
Metonymy	15.79	100.0	0.0
Meronymy	76.32	0.0	28.57
Spatio-temporal function	7.89	0.0	71.43
Other	0.0	0.0	0.0



Distribution of bridging relations (DE)



Distribution of bridging across genres (DE)



Coreference & bridging

- \cdot 17% bridging markables that start a coreference chain
- -> bridging entities are not as important on their own in the text
- 56% coreference chains that have bridging markables connected to them
- -> bridging markables are important for coreference entities



Length of identity chains and number of their bridging markables



Coreference chain length



Bridging distance

- anaphora+cataphora: 20.55 tokens (av. sentence length = 24.87 tokens)
- cataphora: -3.6 tokens
- anaphora: 30.96 tokens
- · distance does not correlate with prominence



Transfer

- looking at German, we annotated English and Russian
- 44% of the German markables transferred
- -> newswire was the most problematic genre



Transfer

- [Die Terroranschläge in Mumbai im letzten Monat] sollten nicht nur die Wirtschaft und das Sicherheitsgefühl Indiens treffen. <...> [Die Täter] haben weder ihre Gesichter verhüllt noch sich selbst in der Manier von Selbstmordattentätern in die Luft gesprengt.
- [Last month's terrorist assault in Mumbai] targeted not only India's economy and sense of security. <...> [The attackers] did not hide their faces or blow themselves up with suicide jackets.



Transfer

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Russian?

• Strategy: Genitive test

Daisy was in [the office] when someone knocked on [the door].

√ [the door] == [the door of the office]



-> Outcomes III

- \cdot a typology of bridging relations
- annotation of bridging with high inter-annotator reliability in 3 languages and 3 text domains
- \cdot near-identity: application to German
- strong correlation between bridging and coreference

Conclusions



- manually annotated multilingual parallel coreference corpus including near-identity and bridging
- applied a knowledge-lean projection approach & manually transferred bridging pairs
- our projection results are competitive as compared to recent work
- our approach is generalisable to other languages and datasets



Future work

- Corpus extension & refining the typology of relations
- Multi-source annotation projection
- Annotated data & guidelines will be available in summer 2017:

ang-cl.ling.uni-potsdam.de/resources



thank you!



References

- Agić, Željko, Dirk Hovy, and Anders Søgaard. "If all you have is a bit of the Bible: Learning POS taggers for truly low-resource languages." The 53rd Annual Meeting of the Association for Computational Linguistics and the 7th International Joint Conference of the Asian Federation of Natural Language Processing (ACL-IJCNLP 2015). 2015.
- Agić, Željko, et al. "Multilingual projection for parsing truly low-resource languages." Transactions of the Association for Computational Linguistics 4 (2016): 301-312
- Clark, Herbert H. "Bridging." Proceedings of the 1975 workshop on Theoretical issues in natural language processing. Association for Computational Linguistics, 1975.
- Ehrmann, Maud, Marco Turchi, and Ralf Steinberger. "Building a Multilingual Named Entity-Annotated Corpus Using Annotation Projection." RANLP. 2011.
- Gardent, Claire, Helene Manuelian, and Eric Kow. "Which bridges for bridging definite descriptions?." 4th International Workshop on Linguistically Interpreted Corpora-LINC'03. 2003.
- Grishina, Yulia and Manfred Stede. "Knowledge-lean projection of coreference chains across languages". Proceedings of the 8th Workshop on Building and Using Comparable Corpora, Beijing, China, page 14. Association for Computational Linguistics, 2015.
- Guillou, Liane, et al. "ParCor 1.0: A parallel pronoun-coreference corpus to support statistical MT." 9th International Conference on Language Resources and Evaluation (LREC), MAY 26-31, 2014, Reykjavik, ICELAND. European Language Resources Association, 2014.
- Hovy, Eduard, et al. "OntoNotes: the 90% solution." Proceedings of the human language technology conference of the NAACL, Companion Volume: Short Papers. Association for Computational Linguistics, 2006.
- Krasavina, Olga, and Christian Chiarcos. "PoCoS: Potsdam coreference scheme." Proceedings of the Linguistic Annotation Workshop. Association for Computational Linguistics, 2007.
- Kunz, Kerstin Anna. 2010. Variation in English and German Nominal Coreference. A Study of Political Essays. Frankfurt am Main: Peter Lang.
- Markert, Katja, Yufang Hou, and Michael Strube. "Collective classification for fine-grained information status." *Proceedings of the 50th Annual Meeting of the Association for Computational Linguistics: Long Papers-Volume 1.* Association for Computational Linguistics, 2012.
- Müller, Christoph, and Michael Strube. "Multi-level annotation of linguistic data with MMAX2." Corpus technology and language pedagogy: New resources, new tools, new methods 3 (2006): 197-214.
- Nedoluzhko, Anna, et al. "Extended coreferential relations and bridging anaphora in the prague dependency treebank." Proceedings of the 7th Discourse Anaphora and Anaphor Resolution Colloquium (DAARC 2009), Goa, India. 2009.



References

- Nissim, Malvina, et al. "An Annotation Scheme for Information Status in Dialogue." LREC. 2004.
- Padó, Sebastián. "Cross-Lingual Annotation Projection Models for Role-Semantic Information. Saarland University dissertation. Published as Volume 21, Saarbrücken Dissertations in Computational Linguistics and Language Technology." German Research Center for Artificial Intelligence (DFKI) and Saarland University (2007).
- Poesio, Massimo, and Ron Artstein. "Anaphoric Annotation in the ARRAU Corpus." LREC. 2008.
- Poesio, Massimo, et al. "Learning to resolve bridging references." Proceedings of the 42nd Annual Meeting on Association for Computational Linguistics. Association for Computational Linguistics, 2004.
- Postolache, Oana, Dan Cristea, and Constantin Orasan. "Transferring coreference chains through word alignment." Proceedings of LREC-2006. 2006.
- Recasens, Marta, Eduard H. Hovy, and Maria Antònia Martí. "A Typology of Near-Identity Relations for Coreference (NIDENT)." LREC. 2010.
- Recasens, Marta, Maria Antònia Martí, and Constantin Orasan. "Annotating Near-Identity from Coreference Disagreements." *LREC*. 2012.
- Riester, Arndt, David Lorenz, and Nina Seemann. "A Recursive Annotation Scheme for Referential Information Status." LREC. 2010.
- Ritz, Julia, Stefanie Dipper, and Michael Götze. "Annotation of Information Structure: an Evaluation across different Types of Texts." *LREC*. 2008.
- Spreyer, Kathrin. "Does it have to be trees?: Data-driven dependency parsing with incomplete and noisy training data." PhD Thesis. 2011.
- Tiedemann, Jörg. "News from OPUS-A collection of multilingual parallel corpora with tools and interfaces." *Recent advances in natural language processing.* Vol. 5. 2009.
- Tiedemann, Jörg. "Rediscovering Annotation Projection for Cross-Lingual Parser Induction." COLING. 2014.
- Yarowsky, David, Grace Ngai, and Richard Wicentowski. "Inducing multilingual text analysis tools via robust projection across aligned corpora." Proceedings of the first international conference on Human language technology research. Association for Computational Linguistics, 2001.
- Zhekova, Desislava. Towards Multilingual Coreference Resolution. Diss. Bremen, Universität Bremen, Diss., 2013, 2013.