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Targeted manual correction

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Combining annotation and features

Conclusion and perspectives

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Improving the parsing of French coordination through annotation standards and targeted features

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and

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Problem

- Parsing Coordinated Structures (CS) in French
 - **1** J'ai mangé une pomme <u>rouge</u> et <u>mûre</u>.
 - 2 J'ai mangé une <u>pomme</u> rouge et une <u>orange</u>.
 - 3 J'ai mangé une pomme rouge et Georges a <u>bu</u> du thé.
- Additional complexity:
 - Intervening arguments and adjuncts
 - Sometimes containing coordination
 - Various forms of ellipsis
 - CS with 3 or more conjuncts
 - Modifiers shared by two or more conjuncts

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French SPMRL annotation for coordination

1 Je vois Jean, Paul et Marie. (I see John, Paul and Mary)



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Software and data

- Talismane (Urieli, 2013)
 - Transition-based parser
 - Left-to-right linear shift-reduce parsing
 - Open source
 - http://redac.univ-tlse2.fr/talismane.html
- French Treebank (Abeillé et al., 2003)
 - Constituent annotation of the French national newspaper *Le Monde*
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Annotation. Software and Error Classification



Annotation, Software and Error Classification



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Initial error classification Initial f-score for coordination in dev corpus: 84.15%. Out of 240 errors analysed:

- Pos-tag mismatch (30%)
- Preposition mismatch (and other simple parallelism) (8%)
- Annotation errors (24%)
 - of which 60% were correctly analysed by Talismane
- Artefacts of annotation scheme (14%)
 - $3^{\rm rd}$ conjunct attached to $2^{\rm nd}$ instead of $1^{\rm st}$
- Semantics required (e.g. two nouns) (12%)
- Other complex cases (ellipsis, etc.) (12%)

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Initial errors: morphology Only 3 our of 240 cases where morphology can help:

[...] on avait parlé de la présidence des AGF à la place <u>de M. Michel</u>
<u>Albert</u> ou <u>de celle</u> du GAN occupée par M. François Heilbronner.
(... they spoke of the presidency of the AGFs instead <u>of Mr Michel Albert</u>
or of that of the GAN occupied by Mr François Heilbronner.)

2 Le conseil d'administration [...] a opté pour la proposition de reprise <u>faite</u> par Bongrain et <u>rejeté</u> celle de Besnier.

(The board of directors **chose** the takeover proposal <u>made</u> by Bongrain and <u>rejected</u> the one made by Besnier.)

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- Targeting simple parallelism (38% of errors):
 - Pos-tag mismatch
 - Preposition mismatch
- If the $1^{\rm st}$ conjunct is correct, generally the $2^{\rm nd}$ as well
- Features target the $1^{\rm st}$ conjunct
- But to do this, they need to guess ahead at the $2^{\rm nd}$ conjunct with little information available

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Identifying the 2nd conjunct

1 Verb coordination: Il s'agit ici d'un jour normal de la semaine et un inventaire scrupuleux exigerait que l'on prenne également en compte l'offre accrue du mercredi.

(We are <u>dealing</u> here with a normal weekday, and a scupulous inventory would require us to take into account the increased offer on Wednesdays.)

2 Comment phrase: A Lourdes, nous signale notre correspondant Jean-Jacques Rollat, la <u>venue</u> et la <u>circulation</u> des pèlerins ont été très perturbées.

(At Lourdes, **signals** our correspondent Jean-Jacques Rollat, the <u>arrival</u> and <u>circulation</u> of pilgrims **was** considerably disrupted.)

Relative clause: Les émissions d'éveil qui ont fait la richesse des chaînes de service public entre <u>1975</u> et <u>1985</u> ont toutes disparu.
(The discovery programmes which constituted the richness of public channels between 1975 and 1985 have all disappeared.)

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Targeted coordination features

• Pos-tag mismatch: N ADJ and N

- Mismatched prepositions: from N to N and from N
- Pos-tag match: V N and N
- 3 conjunct parallelism: N, N and N
- Parentheses: W1 (W2) and W3

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Targeted features: initial results

dev corpus coordination recall:

dev	targeted correct	targeted error	baseline total			
baseline correct	1250	29	1279			
baseline error	61	403	464			
targeted total	1311	432	1743			
test corpus coordination recall:						
test	targeted correct	targeted error	baseline total			
baseline correct	2496	63	2559			
baseline error	167	694	861			
targeted total	2663	757	3420			

Error count:

- Dev error count (recall): from 464 to 432 (-6.9%)
- Test error count (recall): from 861 to 757 (-12.1%)
- Dev f-score: from 84.34% to 85.52% (-7.54%)
- Test f-score: from 85.16% to 86.97% (-12.20%)

Can we do better?

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Targeted manual error correction

- In original error analysis, 24% annotation errors.
- Iterative targeted feature tuning highlights many errors (unexpected feature results)
- Enables efficient targeted manual error correction:
 - 1,488 for train (out of 21,061 coord relations = 7.07%)
 - 106 for dev (out of 1,743 coord relations = 6.08%)
 - 274 for test (out of 3,420 coord relations = 8.01%)

	dev base	dev fix	test base	test fix
train base	84.34	85.08	85.16	85.54
train fix	83.99	85.75	84.99	86.75

Table : Coordination f-score after targeted manual error correction

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Comparing annotation schemes



1st-conjunct headed (1H)



Conjunction headed (CH)



Previous conjunct headed (PH)

Previous conjunct headed 2 (PH2)

- Option: systematically attach all other punctuation to the previous non-punctuation token (+P)
- Results in six schemes: 1H, 1H+P, CH+P, PH, PH+P, PH2+P

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Comparing annotation schemes: results

Scheme:	1H	1H+P	CH+P	PH	PH+P	PH2+P	
Dev							
Coord f-score	85.75	85.60	73.20	86.68	86.96	89.21	
Coord prec.	99.55	99.55	98.88	99.49	99.49	99.41	
Coord recall	75.31	75.09	58.11	76.79	77.24	80.91	
Test							
Coord f-score	86.75	86.94	73.09	88.20	88.44	90.29	
Coord prec.	99.70	99.52	99.38	99.75	99.50	99.71	
Coord recall	<u>76.78</u>	77.18	57.80	79.04	79.59	82.50	

Table : Comparing CS annotation

Coordination f-score error reduction between 1H and PH2+P:

- 24.28% for dev
- 26.72% for test

Combining with targeted features: gains

8.15

13.58

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	None	Features	Scheme	Both		
Dev : base f-score = 85.75						
Beam 1	0.00	4.28	24.28	29.89		
Beam 2	8.49	9.82	32.14	33.40		
Beam 5	9.82	11.44	35.09	34.95		
Test : base f-score = 86.75						
Beam 1	0.00	12.91	26.72	34.64		

 Table : Coordination f-score error reduction with respect to 1H, baseline features, beam 1

F-score increase from 85.75 dev (86.75 test) to 90.73 dev (91.69 test) Best LAS 92.0 dev (92.0 test), Best UAS 93.7 (93.7 test) But, in terms of speed:

19.02

21.81

28.53

28.98

36.83 37.28

• PH2+P scheme: $\times 0.93$

Beam 2

Beam 5

- Beam 2, 5: ×2, ×5
- Targeted features: $\times 22$

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Final error classification

- Errors: 151 (was 250)
- Pos-tag/preposition mismatch: 36% (was 38%)
- Annotation errors: 11% (was 24%)
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- Pos-tag/preposition mismatch: 36% (was 38%)
- Annotation errors: 11% (was 24%)
- Artefacts of annotation scheme: 5% (was 14%)
- Ellipses: 13% (was 5%)
- Semantics required: 23% (was 12%)

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- All three methods successful and relatively cumulative
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