

# HOW TO MAKE MANUAL ANNOTATION MORE EFFICIENT

ENSEMBLE DEPENDENCY PARSING AS A PREPROCESSING STEP TO OBTAIN HIGH-QUALITY ANNOTATION IN AN EFFICIENT WAY

## TASK

### GOALS

- Creating a syntactically annotated textbook corpus for linguistic research
- This poster: Combining automatic and manual dependency annotation to reduce manual workload

### DEPENDENCY PARSING

1 Das	das	ART ART	_	2 SUBJ	-	-
2 ist	sein	V VFIN	_	0 S	-	-
3 ein	ein	ART ART	_	4 DET	-	-
4 Beispielsatz	Beispielsatz	NN NN	_	2 PRED	-	-
5 .	.	\$. \$.	_	0 ROOT	-	-

Fig.1: Dependency parse CONLL format

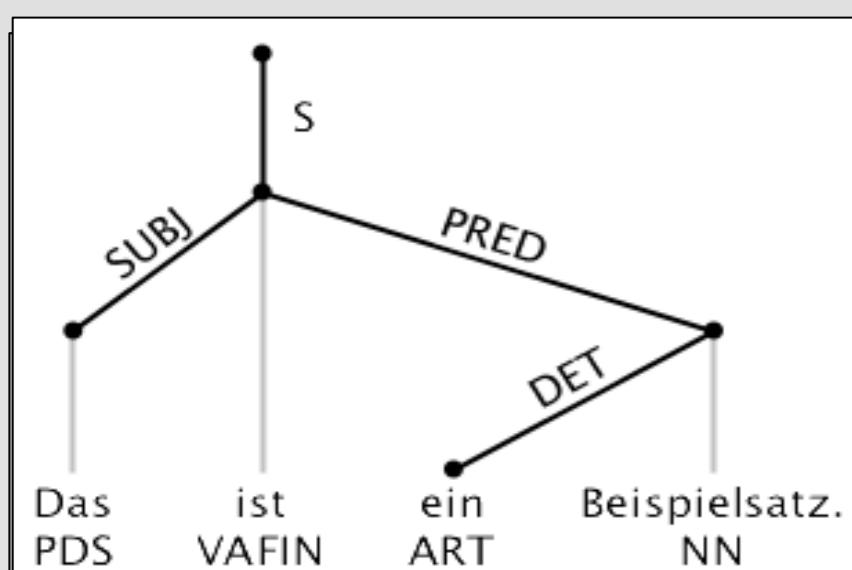


Fig.2: Example dependency parse (HDT AnnoViewer)

### THE “MULTISTRATEGY APPROACH”

- A method of tagging texts by combining the output of a cluster of taggers
- Tagger models result from training different learning algorithms on the same data
- Different taggers create their analyses in different ways such that their errors are uncorrelated
- Hypothesis: A reasonable weighted combination of the tagger choices can obtain better results than the individual taggers do (van Halteren et al. 2001, p. 201)
- Related work: Søgaard (2010), Rehbein et al. (2014)

### RESEARCH QUESTIONS

- What accuracy do the parsers have on the training domain?
- What accuracy do the parsers have on the target domain?
- What accuracy does our ensemble of parsers achieve on the target domain?
- What kind of sentences does the ensemble fail to parse correctly?
- To what extend can the ensemble parser support manual annotation?

### THE TRAINING CORPUS

- Hamburg Dependency Treebank (HDT): Part A
- Contains 101,999 sentences produced by manual annotation and subsequent cross-checking for consistency with DECCA (Dickinson & Meurers 2003; Foth et al. 2014).
- Available free of charge for academic use
- For more information visit: <https://nats-www.informatik.uni-hamburg.de/HDT>

### THE TEST CORPUS

- Texts from the Jena Textbook Corpus (unpublished)
- Three geography textbooks: three double pages
- Average of 54 sentences per text
- Overall 144 sentences parsed

### OUR ENSEMBLE OF PARSERS

- MALT parser (Nivre et al. 2006): Transition-based parsing
- MATE parser (Björkelund et al. 2010): Second order maximum spanning trees-based parsing
- JWCDG parser (The CDG Team 1997-2015): Weighted hand-written rules; developed on the basis of HDT

## QUANTITATIVE RESULTS

### TRAINING ON THE HDT: CROSS-VALIDATION

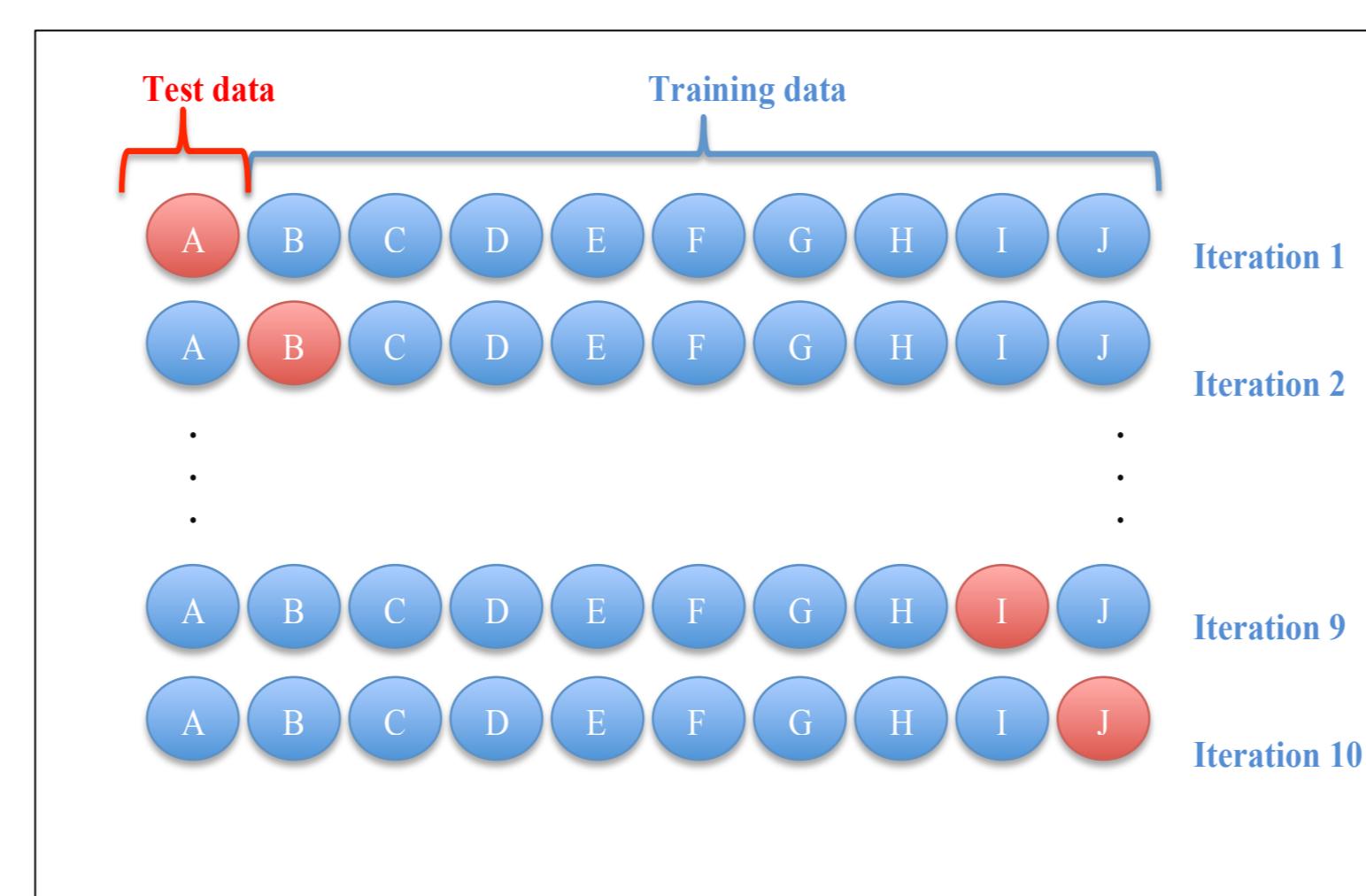


Fig.3: 10-Fold cross-validation

### DEVELOPING A GOLD STANDARD ON THE TEST CORPUS

- Two annotators annotated independently
- IAA: UAS = 0.95 ( $\pm 0.01$ ), LAS = 0.93 ( $\pm 0.01$ )
- $\alpha = 0.93 (\pm 0.02)$  according to Skjærholt (2014)

### ENSEMBLES

ID	Token	Parser-1	Parser-2	Parser-3	Gold	Check
1	Es	2 SUBJ	2 SUBJ	2 EXPL	2 SUBJ	😊
2	ist	0 S	0 S	0 S	0 S	😊
3	ein	4 DET	4 DET	4 DET	4 DET	😊
4	Beispielsatz	2 SUBJ	2 PRED	2 PRED	2 PRED	😊
5	.	0 S	0 S	0 S	0 ROOT	😢

Fig.4: Ensemble creation

Match-3

- Ens-1: Majority vote (Match-3 /-2), default MATE
- Ens-2: Majority vote (Match-3), default MATE except MATE = S or OBJA, then JWCDG

### EVALUATION

- Scores
  - Unlabeled Attachment Score (UAS)
  - Labeled Attachment Score (LAS)
- Settings
  - HDT: Cross validation of 2 statistical parsers
  - Gold Standard: Accuracy of all methods
  - Match-3 instances: Accuracy of parser ensemble

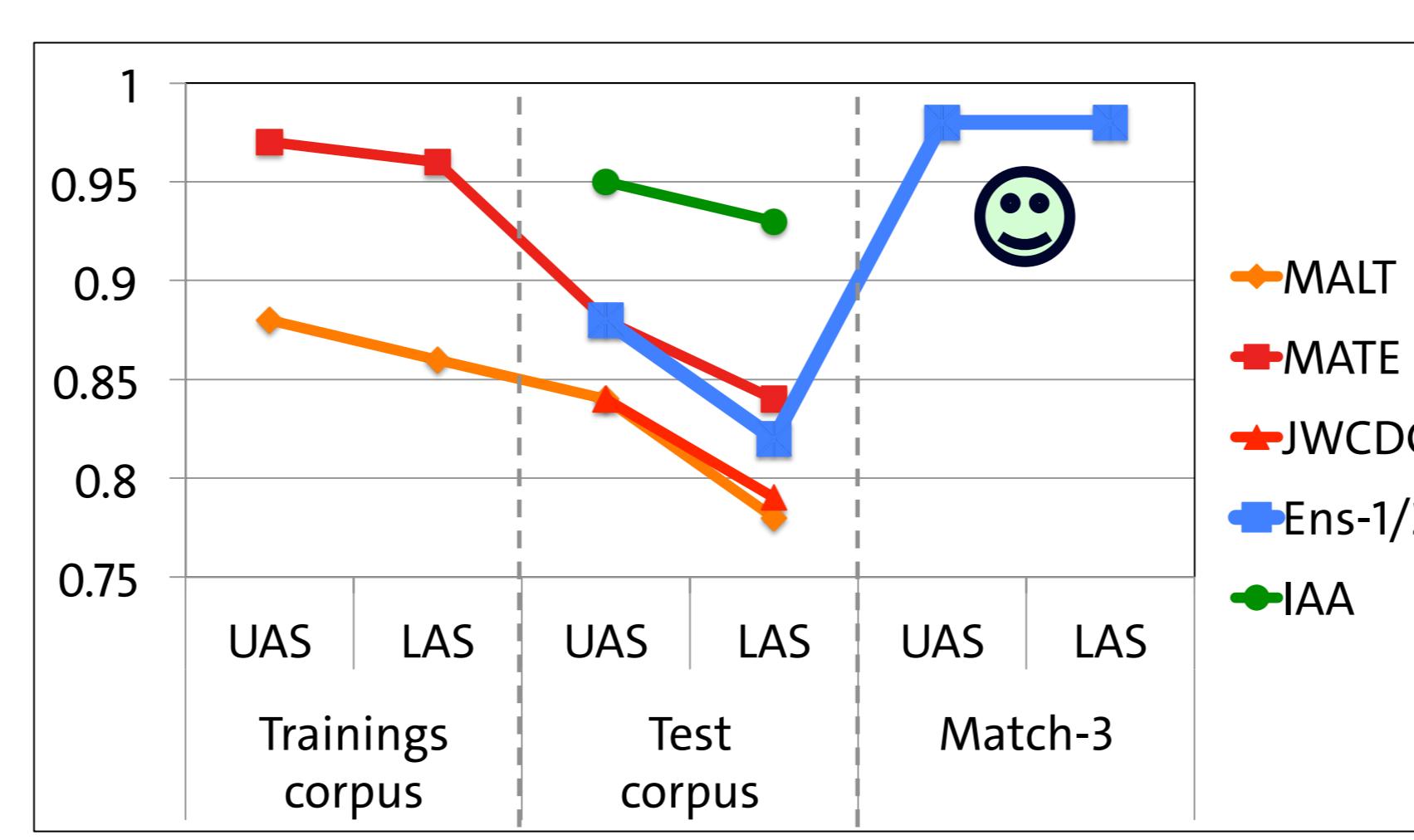


Fig. 5: Accuracy evaluation

- Match-3
  - Ensembles agreed on
    - on 71% ( $\pm 0.10$ ) of all tokens
    - on about 15% of the sentences / text
    - 21 out of 22 agreed-on sentences were correct

## QUALITATIVE RESULTS

### GENERAL CHALLENGES IN DEPENDENCY PARSING

- Valency decision: OBJ vs. PP (depends on training data)
- Attachment ambiguities (semantic decisions)

### PARSER-SPECIFIC CHALLENGES

Parser	Ovgenerates	Comment
JWCDG	APP: 26% (68/262) S: 15% (39/262)	default attachment fragments
MALT	ROOT: 64% (214/335)	incomplete analyses
MATE	S: 13% (30/225) OBJA: 12% (26/225)	fragments confusion SUBJ/OBJA

Table 1

### DOMAIN-SPECIFIC CHALLENGES

- Incomplete sentences (lists, exercises)
  - M4 Auswirkungen des Klimawandels am Beispiel "Starkregen"
  - check-it:
    - Merkmale einer thematischen Karte - hier Bodennutzung - kennen
    - [...]
- Complex coordinations
  - Als praktisch sicher gilt, dass es über den meisten Landflächen wärmere und weniger kalte Tage und Nächte sowie wärmere und häufiger heiße Tage und Nächte geben wird.

## CONCLUSIONS

- Pro: High reliability of ensemble majority vote
- Con: Ensembles do not outperform best individual parser
- Application: Ensemble support of manual parser correction
  - Skipping of perfect labels (AUX, AVZ, DET, GMOD, OBJA, SUBJ) and complete sentence matches
  - Highlighting of OJP and PP
- In addition: Highlighting of lists and exercises
- Future work: Domain adaptation by including the gold standard in the training data

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