





Connection to Literary Studies



The human in the loop
Automatic != fully automatic



Cost Benefit Ratio

Connection to Literary Studies

Ensuring the Satisfaktionsfähigkeit

- Top-Down
 - Operationalize concept of interest
 - Ensure intersubjective application
 - Annotation guidelines (e.g., in a shared task)
 - Corpus, reference data, machine learning, ...
 - Large-scale analysis of texts



Bottom-Up

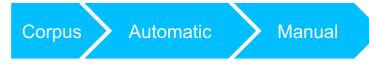
- Select appropriate corpus
- Operationalize concept of interest
 - E.g., instrumental variables / approximate operationalization
- Find patterns (e.g., unsupervised), interpret them
- Connect pattern interpretations to a literary interpretation

Computers

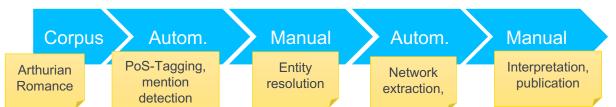
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Automatic != fully automatic

- Full automatization requires (lots of) reference data
- Not available for interesting phenomena
- Interactive tools for exploration



Manual steps can be integrated earlier





Cost Benefit Ratio

Method development

- NLP
 - Large data sets, and constantly new texts
 - (Almost) no upper bound on invested time
- CLS
 - In many cases: No new texts (unless contemporary literature)
 - Paying someone to annotate Goethes Werther might be faster and cheaper than optimizing an NLP model for it
 - Unless: The work on the optimization itself leads to something interesting
- Focus on generic phenomena





Use cases



Context



Black box

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Automatization goals / use cases

- Finding interesting cases ("Finde-Heuristik")
 - Precision matters
 - Interactivity allows users to express/fine-tune their preferences
- Annotation support (candidate generation)
 - Recall matters
 - Practical questions How much annotation do you need before it makes sense? It depends.
- Hypothesis testing
 - "Female characters have a stronger association with illness than male characters, in 19th century novels of genre X."
 - Claims in literary studies do not come in hypothesis form
 - Corpus linguistics toolbox
 - Operationalization crucial
 - Representativity

Context, Ambiguity, Polyvalence

"Peter saw Judy with the binoculars."

- Linguistic ambiguity: Ambiguous sentence, but can be resolved with context
- "Resolved": Readings become implausible
- Claim: Only if assumptions are made
 - that Peter and Judy didn't switch binoculars in between the sentences
 - that binoculars can be used for seeing
 - that ...
- Narratological categories behave similarly
- Ambiguities may remain
- Many ambiguities are unnoticed by humans



Context, Ambiguity, Polyvalence

 Result of text analysis: Propositional content of the text



- Interpretation = Interpretation Theory + Text + Context
- Theory suggests
 - which parts of the text are paid attention to
 - which context to use
- Deterministic, following rules?
- If not, is ruling out interpretations possible?

Black box

- Are black boxes a problem?
 - Depends on use case
 - DHd panel on deep learning: black box OCR ok
 - Empirically validated black boxes are often not a problem, unpredictable performance on new texts is
- Performance on a new text (type) is unknown
 - But we can do empirical validation
 - "Annotate these (representative/difficult) sentences, and we tell you how reliable the pos-tagger was"
- Domain adaptation
 - Users willing to do some annotation can get better results



Conclusions

- Support for semi-automatic processing (integration of manual and automatic processing steps)
- Optimize tools for the right kind of error
 - In general, precision errors are easier to fix than recall errors
- No one knows the performance of a tool on a new text (type)
 - Support empirical validation
 - Users willing to do some annotation can get better results





Vielen Dank!



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