Thesis projects organized by research theme

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NLP Group Meeting

April 10th, 2025



Research interests

- Feedback on every Assessment & Development interview: choose one research topic and focus on that
- Didn't happen so far
- Some of my research interests (in alphabetical order):
 - digital humanities
 - legal NLP
 - low literacy/language proficiency
 - medical NLP
 - sociolinguistics



Today's presentation

Introduction

- A selection of student thesis projects I'm currently involved in
- Grouped by these themes
 - bachelor/master
 - first/second supervisor
 - AI/BI/ADS



Counterpoint for composer detection

- Abe Wolthuis, master AI, 2nd supervisor, proposal finished
- Renaissance and Baroque music uses counterpoint
- Large set of rules on which notes are allowed to be played together
 - Begin and end on either the unison, octave, or fifth
 - ② Use no unisons except at the beginning or end





Composer detection

- Not all composers always followed all rules
- Rule-following patterns as distinctive signature/style of the composer
- Used to detect the composer of unattributed 15th/16th century music pieces
- Thesis project: supervised classifier to identify counterpoint rule usage
- Use as input for composer attribution
- Emphasis on explainability: this composer is predicted because of counterpoint rules a, b, c
- Dataset: 800 compositions by a large number of composers
- Various machine learning models



Semantic data validation

- 3 ADS master students, 2nd supervisor, start block 4
- One student works at UWV to collect unemployment contributions from employers
- Employers submit forms to UWV with financial details
- These forms are checked for conformance with regulations and laws
- UWV keeps a set of rules to do the checks
- Each year these rules need to be updated based on new legislation and interpretation by the tax authority
- Currently manual process



Rule example

Handbook Salary Premiums (from tax authority)

29.30.1 Disability fund (DF)

Within one category and time period you may provide either *increase DF low* or *increase DF high*. The other rubric should be entered as ≤ 0 .

UWV ruleset

rule 2252 IF increase DF high not equal to €0 THEN increase DF low must be €0 rule 2254 IF increase DF low not equal to €0 THEN increase DF high must be €0



Project

- Use information retrieval/text similarity methods to match handbook paragraphs and rules
- 2 Check if rules are obsolete or missing
- Oheck if rules are (still) correct
- Check against applicable law texts (maybe next year)
- Use existing rules in original or modified form to test
- Expert annotations
- Project starts in block 4
- Project also ends in block 4



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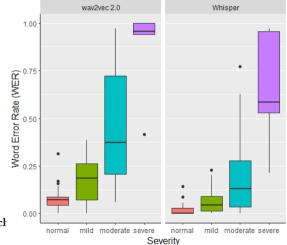
Classifying pathological speech

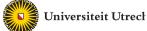
- Luuk Smiesing, bachelor Al, 1st supervisor
- Classify speech disorders in Dutch COPAS dataset
- 8 pathological speech categories
- Wav2Vec2
- Audio SHAP to see which part of the speech explains the predicted class



ASR on the COPAS dataset

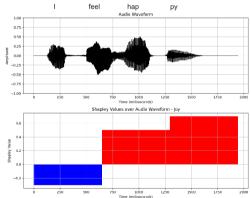
Leo Verspaij 2024 (thesis Radboud Universiteit Nijmegen):





Audio SHAP

Kathleen de Boer 2024 (thesis UU):





Readability

- Michiel Schouten, bachelor Al, 1st supervisor
- Existing tools for increasing readability of text
- Survey: do these tools align with human choices for improving text readability
 - Do people agree with the changes proposed by such tools?
 - 2 What kind of changes would people make themselves?
- Improve tools by not doing types of changes people don't like, implement additional types of changes, apply controversial changes interactively
- Dataset: letters from housing corporations
- Data preparation: run Dutch tool T-Scan and rewrite passages according to T-Scan scores



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Survey questions

- Text presented, randomly either original or rewritten
 - How readable is the given text on a 5 point scale?
 - ② Do you think the text is patronizing?
 - What would you change in the text?
 - (next page) Given the following changes, which would you keep?

Kindly be invited to an orientation meeting on April 12, 2025 at 12:00 with Alice in Our office in Maastricht.

I would like to invite you for a meeting on April 12, 2025 at 12:00 with Alice. The meeting takes place in our office in Maastricht.

 Change types (not shown to participant): active verbs, remove adjectives without clear added value, short sentences
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Preliminary results

- Differences found between acceptance of change types
- Participants (n=20) do not always agree
- Surveys are difficult



Extract medication prescriptions from EHRs

- Maria Dukmak, master AI, 2nd supervisor, proposal phase
- Data and main supervision at UMCU
- Electronic Health Records (EHRs) contain unstructured text with codes and abbreviations
- Medication dosage, duration, usage instructions
 - example usage instructions: 1D1T 3W1WS, 1-2D1T, 1d1t, ???
 - partially standardized
- Goal: transform this into a Common Data Model (CDM) standardized format for use in daily patient care and epidemiological research



Constraints

- Domain-specific Dutch data
- Relatively small, 3M EHRs, 1.8M after cleaning (missing or uninformative values)
- Manual annotation for target labels in CDM
- Data cannot go outside trusted UMCU digital research environment (DRE)
 - Limited computing power
 - Limited internet access
 - Limited pipeline/packages support
- BERT-based models and small generative LLMs

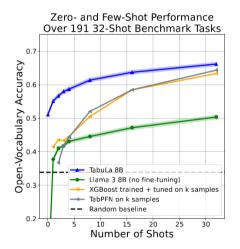


Prediction of treatment outcomes and symptom changes

- Omar Taha, bachelor Al, 1st supervisor
- Collaboration with Amsterdam UMC (data, previous work)
- Predict patient responses to psychiatric treatments
- Dataset: tabular sociodemographic, clinical, neuroimaging data from OCD patients at AMC
- Target variables: predict treatment outcome and symptom changes
- Traditional supervised machine learning models
- Recent models: Tabular Prior-data Fitted Network (TabPFN), TabuLa-8B
 - Tabular foundation models: LLMs pre-trained on tabular data with row-based masking



Tabular LLMs



Gardner et al. 2024, https://arxiv.org/abs/2406.12031

Propaganda detection: entity framing

- Lisa Tanaka, master AI, 2nd supervisor, proposal finished
- **Propaganda**: communication type with the goal of getting people to think or act a certain way
- Framing: representing different aspects and perspectives in a certain way in order to convey latent meaning
 - Paying taxes when a person inherits an estate: calling it "death tax" conveys frame
 of unfair punishment for dying, "estate tax" conveys the frame of justified taxation
 upon obtaining property
- Data: SemEval 2025 task 10 subtask 1 on Entity Framing
- Approach: feature-based and/or generative LLM based



Code Switching

- Nizar Haroun, master Al. 1st supervisor, phase 2
- Code switching: a switch is made between two or more languages, either within a conversation or written text
 - And a scubadiver houdt een schilderij vast
 - Train supervised classifier on small annotated dataset to detect presence of code switching
 - Apply classifier on large non-annotated dataset
 - Identify interesting linguistic or semantic properties of code switched sentences



Approach

- ullet Dataset for training: Kootstra 2010, \sim 1100 sentences collected in controlled lab experiment with NL/EN code switching
- Models: T5, mBart
- Dataset for application: collection of Dutch subreddits
- Linguistic/semantic questions:
 - What is the frequency of code-switching, what are the dialogue patterns?
 - Is the sentiment of code-switched sentences different from monolingual sentences?
 - Which topics are more prominent in code-switched data (using topic modelling, clustering)?
 - ...
- Influence of speaker also interesting but this requires different data



Thesis projects

- Let me know if you are interested to be first or second supervisor on future thesis projects in these topics
- Or to sit in on a meeting or the defense of the current students
- Or if you want to get in touch with any of the partners
 - UWV, UMCU, AMC, ELSA Lab, smaller companies
- Questions?

