732A81/TDDE16 Text Mining (HT2024)

Project kick-off

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Project timeline



- Please use the chance to get individual supervision!
 - 19.+20.12.: Only via Zoom or e-mail
 - 21.12.-6.1.: Only via e-mail (but please don't expect a fast response)

- From 7.1.: Supervision bookings possible again

Project kick-off

Plan for today

1. Formal Requirements

2. Example Projects

- 3. Practical Tips
 - Project Structure
 - Getting Help

4. Your Questions

Project kick-off 2

Formal Requirements



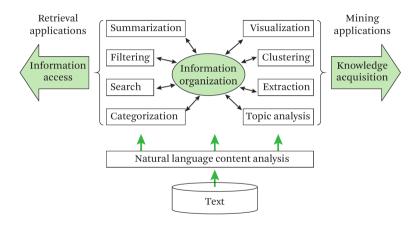
Purpose of the project

The project module primarily tests to what extent you are able to

- 1. identify, formulate and solve problems within the area of text mining, and
- 2. clearly present and discuss the conclusions of a project work.

From the course memo

Is your problem "within the area of text mining"?



Zhai and Massung (2016)

Knowledge requirements

- You identify and formulate a substantial text mining problem with some help from a teacher.
- In working on your problem, you
 - implement and apply suitable text mining methods,
 - analyse experimental results with appropriate evaluation methods, and
 - summarise them with well-developed judgements.
- You clearly present and discuss the conclusions of your work.

https://liu-nlp.ai/text-mining/project/

Submission

- The project report is submitted as a single PDF file via Lisam.
- Additionally, I will prepare a project registration form for you to submit:
 - Your project title & abstract. (can be published as example projects in coming years)
 - A link to your code repository.
 - Either on Github or on LiU's Gitlab server.
 - For private repos, you must give access to mbollmann (Github) or marbo59 (Gitlab).
 - A declaration of whether and how you used generative AI in preparing your report.

Example projects from previous years



Sentiment Analysis of IMDB Movie Reviews

- How well can we predict the sentiment of movie reviews?
 - Used the existing *IMDB* movie reviews dataset
- Trained and evaluated different types of classifiers.
 - Naive Bayes, Support Vector Machine, BERT, RoBERTa
- Performed a manual error analysis of reviews misclassified by BERT/RoBERTa.
 - Gain additional insights that could help improve models further

Explainable Authorship Attribution using BERT

- Can we identify an author from the way they write?
 - Found an *author identification dataset* on Kaggle
- Trained and evaluated different types of classifiers.
 - Naive Bayes, Logistic Regression, BERT
- Which features does the BERT model consider "characteristic" for each author?
 - Used integrated gradients method to extract features that were most important for the BERT classifier when making its predictions

Analyzing how the Refugee Crisis Affected Parliament Speeches using BERTopic

- Did the Syrian refugee crisis cause an increase in immigration-related topics being discussed in the Swedish parliament?
- Analysed the change of topics over time, as modelled by BERTopic.
 - BERT-based topic modelling technique
- Evaluated the performance using a **coherence** metric as well as manual analysis.

Generating free-text explanations with GPT-2/GPT-3

- Natural language inference (NLI) is the task of predicting whether two sentences entail or contradict each other.
 - "An adult holds a stick." contradicts "An adult walks away empty-handed."
- Can we generate free-text explanations for these predictions?
 - "Holding a stick implies using hands, so it is not empty-handed."
- Compared two generative models using different techniques and performed human evaluation on the generations.
 - GPT-2 with fine-tuning νs . GPT-3 with few-shot prompting

Family tree extraction for Tolkien's world

- Can we automatically extract family trees for Tolkien's characters using encyclopedia entries about them?
 - Scraped the Lord of the Rings Wikia site
 - Implies detecting character names and the relations between them
- Evaluated using the "infoboxes" section of each character page as ground truth.

Project abstracts from previous years

• A selection of **project abstracts** from previous years is on the course website.

Note

It is perfectly fine to work on the same topic as a project from previous years!

- Perhaps you can find a better approach, or do a different analysis, or...
- You are unlikely to have *exactly* the same ideas as another student.

Practical Tips



A common project structure

- 1. Identify your problem (ca. 8 hours)
- 2. Design your approach (ca. 32 hours)
- 3. Evaluate your approach (ca. 32 hours)
- 4. Produce your report (ca. 16 hours)

Phase 1: Identify your problem

- Is there a specific task you want to work on?
 - "relation extraction to construct family trees of Tolkien's characters"
- Is there a (limited-scale) research question you want to answer?
 - "Did the Syrian refugee crisis cause an increase in immigration-related topics?"
- One way to get inspiration is to look for datasets first.
 - Kaggle, HuggingFace Datasets, Riksdagens öppna data, many "shared tasks" such as RepEval 2017, ...

Phase 2: Design your approach

- 1. Select one or more datasets.
- 2. Write code to **process** the data and prepare models/experiments.
 - Use existing libraries, e.g.
 ☐ spaCy, ☐ Gensim, ☐ HuggingFace Transformers, ...
 - No requirement on the programming language!
- 3. Choose a method for **evaluating** your results.
 - Consider both evaluation *measures* as well as *baselines*.
 - Think about this before implementing your approach!

Phase 2: Design your approach – tips & tricks (I)

- It's a good idea to review previous work related to your problem.

 - Search via Semantic Scholar
- This can give you new ideas as well as pointers to existing code and datasets!
 - Z Papers with Code specifically collects academic papers with code implementations.
- Collect references as you go so you can cite your sources appropriately!
 - Includes code that you re-used.

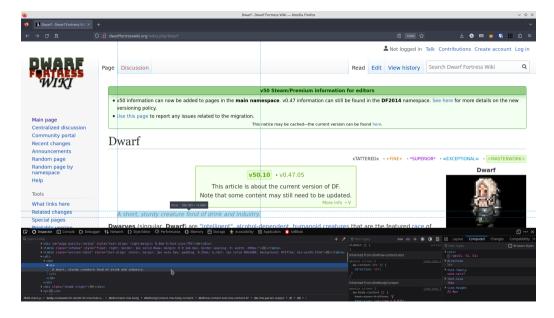
Phase 2: Design your approach – tips & tricks (II)

If your project uses machine learning in any way, consider:

- How to avoid machine learning pitfalls, by Michael A. Lones
 - Before you start to build models
 - How to reliably build models
 - How to robustly evaluate models
 - How to compare models fairly
 - How to report your results

Phase 2: Design your approach – tips & tricks (III)

- Web scraping is the process of automatically extracting data from websites.
- Involves fetching a web page, parsing it, and finally extracting data from it.
 - BeautifulSoup
 - − Scrapy
 - Trafilatura
- May violate the terms of use of some websites and/or constitute copyright infringement.
 - Countermeasure include blocking the scraper's IP address.



Web scraping example: Trafilatura

```
1 >>> from trafilatura import fetch_url, extract
2 >>> downloaded = fetch_url("http://dwarffortresswiki.org/index.php/Dwarf")
3 >>> text = extract(downloaded, include_tables=False)
4 >>> print(text)
5 - A short, sturdy creature fond of drink and industry.
6 This is a masterfully-designed engraving of a Dwarf and a battle axe.
7 Dwarves (singular, Dwarf) are "intelligent", alcohol-dependent, humanoid creatures that are the featured race of fortress mode, as well as ...
```

Phase 3: Evaluate your approach

- Run your experiments and interpret your results.
 - Quantitative measures, e.g., accuracy, precision, recall, topic coherence, perplexity, ...
 - Qualitative analysis, e.g., human evaluation, manual inspection of interesting test cases, ...
- Remember that most evaluation methods are relative!
 - Can't conclude that "X works well", but "X works better or worse than Y"
- Consider Colab or Maggle if you need computing resources.

Phase 4: Produce your report

- Use the required template and revisit the suggested structure.
- Consider drafting sections already during the previous phases!
 - e.g. Data & Method during phase 2
- Consider how to present your project in a clear and understandable way.
 - e.g. giving examples of data or inputs/outputs, summarizing important figures in a table,
 visualizing complex concepts in a figure, visualizing results in a graph/plot, ...

How to get help

- Pitch your project idea to me!
- I am offering one-on-one meetings for you to get feedback.
 - Book a time slot through the link on the website
- I am also available via e-mail.
 - Use whichever option works best for you and the question(s) you have!

Your Questions



66 How can we find a project idea?

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- Check the project abstracts on the course website
- Check the example projects in this presentation
- Check slide 15

66 Can we use methods or topics from outside the course?

"

• Yes! As long as it fits the framework for text mining and does natural language analysis in some way, you can work on any topic you want.

Are we restricted to certain packages or frameworks?

"

• No! You can use any tools, libraries, frameworks, and programming languages you want.

66 How much time to spend on coding vs. report writing?

That's a difficult question to answer in general. See slide 14 for a rough estimate.
 Don't underestimate the report writing part, though, and ideally start drafting early!

How much data do we need? What if data collection takes a significant amount of time?

- You need enough data to draw meaningful conclusions. How much that is depends on what you want to do with it; ask me if unsure! There's certainly no need to work with huge datasets of millions of documents, though.
- Spending effort on gathering data can contribute to a higher grade for your project.

44 Are there any good web-scraping techniques?

"

• Yes! See the pointers on slide 19.

What is the length expectation of the paper?

"

• 3–5 pages of content in the ACL template. Please check the project report formalities for more details!

What is the minimum amount of sources we should use?

- · There is no minimum requirement.
- For a passing grade, make sure to cite all sources of text, data, or ideas that you use in your project.
- For a higher grade, you need to cite scientific articles, but quality is more important than quantity. Citing one paper and connecting it really well to your own work is better than citing ten papers without any meaningful discussion.

Does the project have to have scientific novelty?

"

· No!

What are the points that you specifically check for the highest grade?

"

- Please check the project report formalities for more details on the assessment criteria.
 For the highest grade, your report should be very good in each of the three aspects mentioned there.
 - But keep in mind that it's not a checklist, and you don't need to achieve everything there
 to get the highest grade.
- If unsure, ask! Tell me that you are aiming for a higher grade and what your plans are to achieve this, and I can give you feedback on how well your plans fit the criteria.

