- Motivation
- Prompt Engineering Project
- GPT-3 Playground
- Examples
- Strategies

What is Prompt Engineering about?

- Using existing pre-trained language models to solve a broad range of tasks.
  Might be classification tasks or generative tasks
  (where the task is to generate an output sequence).
- Justification: Language models generalize well and can be applied to tasks which they were not trained on.
  - Just need to ask the right way!
- Might allow for tasks that were not possible before
- Benefits: No (re-)training, no finetuning
- Behaviour of the model is almost completely controlled by the model input (i.e. the prompt).
  - Must optimize the input prompt to receive optimal results (prompt engineering).

#### Disclaimer

- May be seen as orthogonal to the approaches that you are used to Doesn't replace the classic DL approaches that we studied before
- Many aspects stay the same
  - Careful problem analysis
  - Data aquisition and exploration
  - Rigorous and metrics-based evaluation
- May seem like alchemy

But didn't Feature Engineering and DL Model Engineering feel like alchemy at first, too?

#### Disclaimer

- New problems arise
  - Extremely hard to detect train-test leakage
    Using canary strings in benchmark tasks
  - Non-rigorous evaluation leads to wrong beliefs about performance
  - Model availability and computation cost
  - Blackbox models (even more extreme than in classic DL!)
  - Undetected bias

#### Relevance to term projects

- Provides a way to quickly demonstrate language-based tasks
  Without actually designing and training a model
- Forces you to think about model inputs and outputs
- Helps to visualize the text data in the single processing steps Actually see what is happening!
- Might actually be used in the term projects for some pipeline steps
  E.g. for refining the data
- Next week, we will talk about making prompt engineering accessible to software through APIs

Large Language Models (LLM)

- Transformer-based language models
- Typically billions of parameters
- Built to generate an output sequence based on an input sequence
- Many publicly available models
  - e.g. Open Pre-trained Transformers (OPT)
- Some only through APIs
  - e.g. Generative Pre-trained Transformer 3 (GPT-3)

## **Prompt Engineering Project**

Workflow

One mini project for each student.

- Find an interesting prompt task
  - i.e. a question that should be solved using prompt engineering Examples follow. (Please do not choose exactly the tasks from the lab.)
- 2. To make sure that you do engineering and not just running some arbitrary prompt:
  - Choose an interesting aspect to investigate further e.g. How well do different prompt engineering strategies work?
  - If the quality of the results can be measured objectively, a plot would be useful

### **Prompt Engineering Project**

Deliverable: Slides

To present the results to your fellow students, please prepare exactly 3 slides:

- 1. Short title of the project + short problem description in max 2 sentences
- 2. Example prompt that aims to solve the problem + output also specify the used model
- 3. Plot or chart or ... to document your engineering investigation

Please send the slides as a PDF to niklas.deckers@uni-leipzig.de until 12.06.2022, 22:00,

and prepare for a 2-minute presentation in class.

**GPT-3 Playground** 

https://beta.openai.com/playground/

- Basic usage
- Engine selection
- Stop sequences

#### Basic examples

- □ Q&A
- Magic Spells
- Summarize for a 2nd grader
- Analogy extraction
- □ Group Name Generation

#### More examples

```
□ https://beta.openai.com/examples
```

https://github.com/google/BIG-bench/tree/main/bigbench/ benchmark\_tasks#readme

More ideas related to web data

- Argument generation
- Generating replies to reddit posts from a specific subreddit
  r/explainlikeimfive, r/AmItheAsshole, r/changemyview, r/WrongAnswersOnly
  ...
- Generating/understanding/explaining humour
- Your own ideas?

#### General advice

- Do not think that prompts should "command the machine" (imperatives)
- Do not think that "the machine is talking to you"
- Rather think about the output as the "most probable completion" Depends on the data the LLM had been trained on, beware of internet language!
- □ This might also lead to mere reproduction of training data
- Classification tasks are also possible
  Computing which of the given options is more probable to be the completion

PE:I-13 Prompt Engineering

Defining context/identity/intent

- Defining context allows the LLM to adjust to a specific task
- Specifying the intent allows to e.g. prevent insulting language
- May significantly improve result quality

#### Few-shot learning

- Idea: The LLM has been trained on a too broad scope, so provide it with successful examples
- Also useful to enforce a certain output format
- Cover a broad range of samples for your task
- Samples must have high quality
- Remember: Computation cost increases linearly

#### Instructions

- Originally, LLMs did not perform well with instructions This aligns with our expectations from the concept of the "most probable completion"
- However, modern LLMs are often optimized to follow specified instructions
- Might help to specify/clarify the task explicitly

Intermediate steps

- Splitting the task up into multiple intermediate steps
- May be represented by multiple execution steps
- Should also be reflected in the few-shot learning samples

Allowing and expecting edge cases

- Allow for special values as a result (e.g. exception statements)
  Can be specified in the context/instruction and included in few-shot learning samples
- For classification-like tasks, cover a broad scope of classes (e.g. neutral)
- Expect outputs that don't match your desired pattern