

Prompt engineering

Date _____

Subject _____

• LLM generates text, how??

- we give LLM a prompt (a sequence of tokens)
- LLM generates text one token at a time.
- each new token is generated conditioned on:
 - the original prompt
 - all tokens generated so far

• next-token prediction Formula

- at generation step i , the model computes:
 $p(w_i | w_{i < i})$
- the model outputs a distribution over a vocabulary tokens
- we choose the next token by:
 - ☐ sampling, or ☐ Greedy choice, or ☐ beam search
- this repeats until:
 - ☐ "end of text" token appears
 - ☐ a preset length is reached

- **prompt:** a text string that user provides to instruct the LLM.
- **purpose:**
 - tells the model what task to perform
 - provides context, style, target format, constraints
- **Prompt engineering:**
 - the process of designing prompts that makes the model perform a task well.
 - **Good prompts:**
 - clearly describe ~~target~~ the task
 - specify only constraints (format, style, length)
- **Few-shot vs zero-shot prompting**
 - **Zero-shot prompting:**
 - Prompts contains instructions but no examples
 - example:
"translate into french"
 - **Few-shot prompting:**
 - we include labeled examples in the prompt
 - Helps the model understand the pattern or task.

Demonstrations in Prompting:

- Demonstrations: small labelled examples inserted into the prompt, usually drawn from a labeled training set.

How demonstrations are chosen.

- sometimes manually selected by humans
- sometimes chosen automatically
 - an optimizer searches for the set of demo examples that gives the best performance.

number of demonstrations:

- only a few examples usually needed (few-shot)
- adding more gives diminishing returns.
- too many can cause:
 - overfitting to specific examples
 - worse generalization

what demonstrations really achieve??

- main purpose: show the task structure.
- They do not need to show perfectly. even demonstrations with wrong answers can still help the model.

- Definition:

In-context learning is when a language model improves its behaviour on a task just by being shown examples in the prompt, not by updating its weights.

- the learning happens on the fly, inside model's temporary context window.

- the learning that happens during prompting, improves the model's performance without :

- gradient descent
- weight updates
- retraining