

• 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:
$$p(w_i | w_{1:i})$$
- 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 ~~what~~ 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.
 - examples:
english → persian
dog → سگ
house → خانه

- **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
 - sometime 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