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Prompting

- **What is Prompting?**

- Prompting refers to the practice of crafting specific input instructions or queries (called prompts) to guide a language model like GPT to perform a desired task.
- Unlike traditional machine learning models, which require fine-tuning on task-specific data, prompting leverages the model's pre-trained knowledge to solve tasks directly during inference.

- **How Prompting Works?**

- Prompting involves presenting the model with:
 1. Instructions: Explain the task in natural language.
 2. Context (optional): Provide background information or examples to establish the task.
 3. Query: The specific input or problem to solve.
- The model processes the prompt and generates a response based on the provided information.

Types of Prompting: Zero-Shot Learning

1. Zero-Shot Learning

- **Definition:** The model is asked to perform a task with *no prior examples* of input-output pairs provided in the prompt.
- **How It Works:**
 - The task is defined purely by the instructions or description in the prompt.
- *Example:*

Prompt: “Translate this sentence to French: 'I love programming.’”

Expected Output: “J'aime la programmation.”

- **Challenge:** Requires the model to generalize entirely from its pre-trained knowledge without specific task examples.

Types of Prompting: One-Shot Learning

2. One-Shot Learning

- **Definition:** The model is given one example of an input-output pair in the prompt before being asked to perform the task on a new input.
- **How It Works:**
 - The single example serves as a hint or demonstration of the task.
- *Example:*

Prompt: “Use the example to translate the statement to French.

Example: ‘I love programming.’ -> ‘J'aime la programmation.’

Statement: ‘I enjoy learning.’”

Expected Output: “J'aime apprendre”

- **Challenge:** The model must infer the task and generalize correctly from a single example.

Types of Prompting: Few-Shot Learning

3. Few-Shot Learning (also known as in-context learning)

- **Definition:** The model is provided with a few examples (typically 2-10) of input-output pairs as part of the prompt.
- **How It Works:**
 - These examples establish a pattern or context for the task.
- *Example:*

Prompt: “Use the examples to translate the statement to French.

Example 1: ‘I love programming.’ -> ‘J'aime la programmation.’

Example 2: ‘I like coffee.’ -> ‘J'aime le café.’

Statement: ‘I enjoy learning.’”

Expected Output: “J'aime apprendre”

- **Challenge:** Balancing the number of examples to provide sufficient guidance while keeping the prompt within token limits.
- **Advantage:** Often yields better performance than zero- or one-shot by reducing ambiguity.

Types of Prompting: Chain-of-thought

4. Chain-of-thought prompting

- **Definition:** The prompt explicitly encourages the model to generate step-by-step reasoning to arrive at the answer.
- **How It Works:**
 - This approach mirrors human problem-solving, where we often break tasks into logical intermediate steps to arrive at a conclusion.
 - *Example:*

Prompt: “Solve the following problem step by step:
If a train travels 60 miles in one hour, how far does it travel in 3 hours?”

Expected Output: “The train travels 60 miles in 1 hour. In 3 hours, it travels $60 \times 3 = 180$ miles.
The answer is 180 miles.”

- Advantage: Combining CoT with few-shot or zero-shot sometimes yields better performance than the traditional settings.

Prompt Engineering: Foundational Work

- Brown, T., Mann, B., Ryder, N., Subbiah, M., Kaplan, J. D., Dhariwal, P., ... & Amodei, D. (2020). [*Language models are few-shot learners.*](#) Advances in neural information processing systems, 33, 1877-1901.
- Wei, J., Wang, X., Schuurmans, D., Bosma, M., Xia, F., Chi, E., ... & Zhou, D. (2022). [*Chain-of-thought prompting elicits reasoning in large language models.*](#) Advances in Neural Information Processing Systems, 35, 24824-24837.
- Kojima, T., Gu, S. S., Reid, M., Matsuo, Y., & Iwasawa, Y. (2022). [*Large language models are zero-shot reasoners.*](#) Advances in neural information processing systems, 35, 22199-22213.
- Min, S., Lyu, X., Holtzman, A., Artetxe, M., Lewis, M., Hajishirzi, H., & Zettlemoyer, L. (2022). [*Rethinking the role of demonstrations: What makes in-context learning work?*](#) arXiv preprint arXiv:2202.12837.
- Madaan, A., & Yazdanbakhsh, A. (2022). [*Text and patterns: For effective chain of thought, it takes two to tango.*](#) arXiv preprint arXiv:2209.07686.

Should I do Prompting?

- **Why Prompting?**

1. No Fine-Tuning Needed: Models can perform tasks directly with prompts, reducing the need for task-specific training.
2. Task Versatility: Prompts can be used for translation, summarization, question-answering, creative writing, and more.
3. User-Friendly: Natural language makes prompt design accessible to *non-experts*.

- **Challenges of Prompting**

1. Ambiguity: Poorly designed prompts can confuse the model, leading to incorrect or irrelevant outputs.
2. Length Limits: Prompts must fit within the model's maximum token limit, requiring concise yet effective design.
3. Sensitivity: Minor changes in wording or structure can significantly affect performance.
4. Context Dependence: Some tasks require substantial context to produce reliable outputs.

Tips for Prompting

- [Hugging Face](#) documentation.
- [OpenAI](#) documentation: Defines 6 strategies
- Some tips
 1. **Be Explicit:** Clearly state the task and desired output format. Example: "Write a short summary in bullet points."
 2. **Use Demonstrations:** Include high-quality examples to guide the model.
 3. **Leverage Chain-of-Thought:** Encourage reasoning for complex tasks.
 4. **Iterate and Test:** Refine the prompt based on the output quality.
 5. **Avoid Ambiguity:** Make instructions precise to reduce misinterpretation.

Thank You

