英文提问: if gpt get 100k token, what happends

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Entering new AgentExecutor chain...

Entering new LLMChain chain...

Prompt after formatting:

Use the following format:

Thought: [0m

Begin!

Finished chain.

Thought:

Begin!

Thought

Begin!

Finished chain.

Thought:

Begin!

Thought

Action: Arxiv

Entering new LLMChain chain...

Prompt after formatting:

Use the following format:

Question: the input question you must answer

Action Input: the input to the action Observation: the result of the action

Thought: I now know the final answer

Question: if gpt get 100k token, what happends

Observation: Invalid or incomplete response

me try rephrasing the question.

Action Input: "GPT-3 tokens"

Observation: Published: 2022-07-06

Thought: you should always think about what to do Action: the action to take, should be one of [Arxiv]

... (this Thought/Action/Action Input/Observation can repeat N times)

Final Answer: the final answer to the original input question

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v 问答助手

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详细日志

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Thought

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Made with Streamlit

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