

Survey: Research on "Chain-of-Thought (CoT)" Quality in Large Language Model (LLM) Code Generation

Survey Instructions: This survey aims to investigate the quality issues related to CoT encountered by software developers when using reasoning-capable LLMs (such as OpenAI o1, DeepSeek-R1, etc.) for coding assistance. Your feedback will be used for academic research to help improve the reliability of code generation by LLMs. This survey is anonymous and takes about **3-5 minutes** to complete.

Part 1: Background Information

1. **Your software development experience is approximately:**
 - ☐ Less than 1 year
 - ☐ 1 - 3 years
 - ☐ 3 - 5 years
 - ☐ More than 5 years
2. **Which reasoning-supported LLMs do you typically use in programming tasks? (Select all that apply)**
 - ☐ OpenAI o1 series (o1-preview, o1-mini) [
 - ☐ DeepSeek-R1
 - ☐ Gemini series (Gemini 2.0 Flash Thinking, Gemini 2.5 pro, Gemini 3 pro, etc.)
 - ☐ Claude series (Claude 3.5 Sonnet, Claude 3.7 Sonnet, etc.)
 - ☐ Qwen3 series (Qwen3-Max-Thinking-Preview, etc.)
 - ☐ Other: _____
3. **When using LLMs to solve complex programming problems, do you read the "Chain-of-Thought/Reasoning Process" generated by the LLM?**
 - ☐ Never (I only look at the final generated code) **(Please stop here, thank you for your participation)**
 - ☐ Occasionally
 - ☐ Frequently
 - ☐ Always

Part 2: Verification of CoT Quality Factors

In our research, we have summarized **5 categories of common CoT quality issues**. Please rate the frequency with which you encounter these issues based on your actual experience.

(**Note:** Please refer to "Figure 1: Taxonomy of factors influencing CoT quality" for the classification structure)

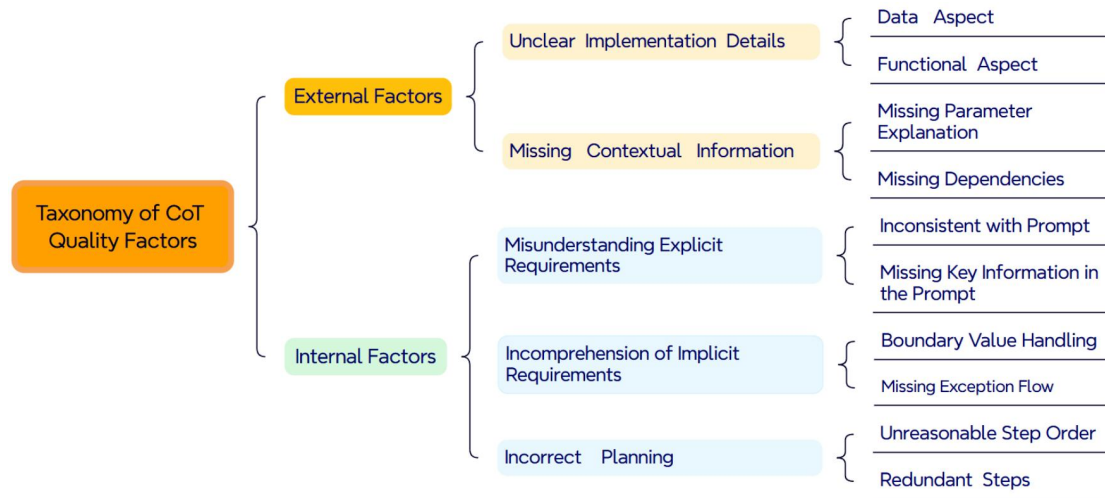


Fig. 1: Taxonomy of factors influencing CoT quality

【External Factors】

4. Unclear Implementation Details

✧ Data aspect

- **Description:** Due to a lack of explicit constraints on data structure or data types in the Prompt, the model produces logical deviations during the generation of reasoning steps. *Example:* You asked to "process a set of data" but didn't specify if it was a List or a Set. The CoT proceeded with logic for a List when it should have been a Set.

- **Have you encountered similar situations?**

- ☐ Never encountered
- ☐ Occasionally encountered
- ☐ Frequently encountered
- ☐ Extremely frequent

✧ Functional aspect

- **Description:** The Prompt's description of functionality is not precise enough, causing the model to misunderstand the operational logic. *Example:* You asked to "remove separators" but didn't specify "remove all" or "only remove trailing ones." The CoT incorrectly assumed "remove all."

- **Have you encountered similar situations?**

- ☐ Never encountered
- ☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

5. Missing Contextual Information

✧ Missing Parameter Explanation

- **Description:** The prompt lacks explanations for input parameters in the function signature, causing errors during CoT generation. *Example:* A function has a parameter named flag. The CoT assumed it was a control switch (boolean), but it was actually a bitmask.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

✧ Missing Dependencies

- **Description:** The task depends on certain external libraries or specific variable names not provided in the Prompt, leading to hallucinations of non-existent variables or dependencies in the reasoning steps. *Example:* The CoT fabricated a function named util.check_valid() to assist with validation during reasoning, but this function does not exist.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

【Internal Factors】

6. Misunderstanding Explicit Requirements

✧ Missing Key Information in the Prompt

- **Description:** The Prompt clearly stated a key condition, but the CoT failed to extract it during reasoning. *Example:* The Prompt explicitly stated ignore case, but the CoT's reasoning steps did not mention this at all and processed it as case-sensitive.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

✧ **Inconsistent with Prompt**

- **Description:** The Prompt explicitly stated a requirement, but the reasoning steps conflicted with that requirement.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

7. Incomprehension of Implicit Requirements

✧ **Boundary Value Handling**

- **Description:** The CoT only planned for normal cases and ignored boundary cases.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

✧ **Missing Exception Flow**

- **Description:** The CoT did not consider potential error scenarios and lacked robustness in design.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

8. Incorrect Planning

✧ **Redundant Steps**

- **Description:** The CoT contains completely unnecessary or redundant steps.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

✧ **Unreasonable Step Order**

- **Description:** The order of steps in the CoT is incorrect.

- **Have you encountered similar situations?**

☐ Never encountered

☐ Occasionally encountered

☐ Frequently encountered

☐ Extremely frequent

Part 3: Impact and Feedback

9. In your experience, when the CoT exhibits the logical defects mentioned above, the final generated code is usually:

☐ Always incorrect (If CoT is wrong, the code is definitely wrong)

☐ Frequently incorrect

☐ Not necessarily (Often the CoT logic is messy, but the model writes the correct code from memory) ☐ Not really related

10. In your opinion, the capability most lacking in LLMs during code generation is:

☐ Proactive Clarification (Ability to clarify ambiguous requirements)

☐ Planning (Complex logical planning ability)

☐ Robustness (Consideration of boundary/exception cases)

☐ Self-Correction (Ability to correct its own errors)

11. Have you encountered other types of reasoning errors? (If yes, please briefly describe)