

Hell or High Water: Evaluating Agentic Recovery from External Failures

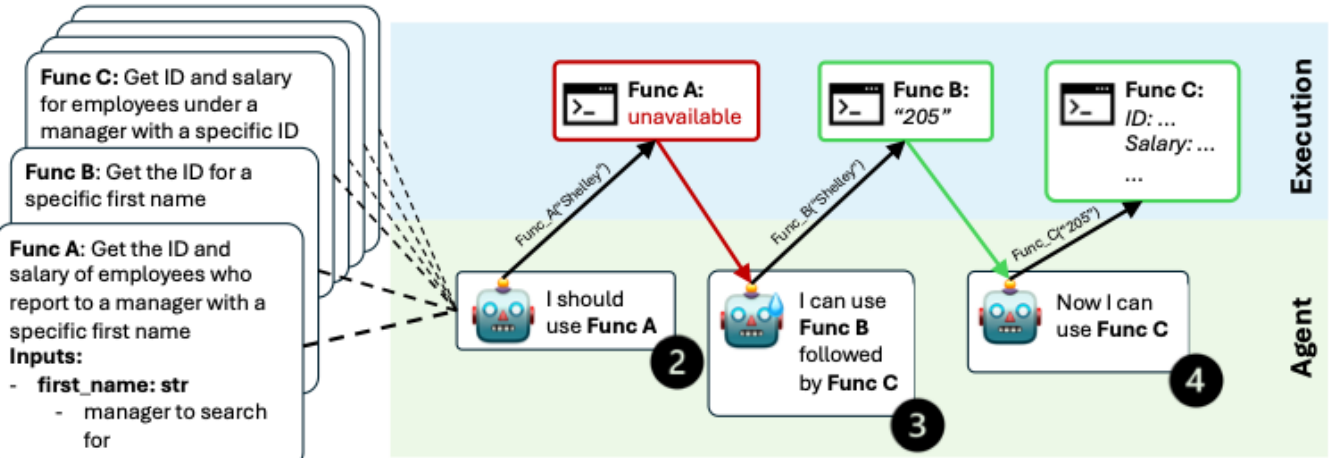
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Tools break in the real world — how well do LLMs deal with tool failures?

Agentic tool use research has focused on analyzing and correcting **agent errors**. A few works study whether LLMs can *identify tool errors*. Our work goes one step further and measures how well LLM agents find backup solutions if tools fail.

1 What are the employee ids of employees who report to Shelley, and what are their salaries?



Large set of functions + docs

Agent Workflow

We introduce HOHW, a tool-use benchmark where problems remain solvable even when tools break adversarially. HOHW consists of 830 problems and 4450 available tools

- The agent correctly tries to use Func A, but encounters an external error outside its control: Func A is unavailable. Therefore, it must form a backup plan to use the composition of Func B and Func C.

SQL Query: is sampled from Spider

```
SELECT employee_id, salary
FROM employees WHERE
manager_id = (SELECT
employee_id FROM employees
WHERE first_name = 'Payam')
```

Parameterized SQL Query

```
SELECT employee_id, salary
FROM employees WHERE
manager_id = (SELECT
employee_id FROM employees
WHERE first_name = ?)
```

Decompose to Subqueries

```
SELECT employee_id
FROM employees
WHERE first_name = ?
```

```
SELECT employee_id,
salary FROM employees
WHERE manager_id =
(SELECT * FROM
scratchpad)
```

Process:

Input: text-to-SQL (Spider)
Output: task + python tools

- Find SQL with subqueries
- Wrap subqueries individually (func. set 2)
- Wrap original query (func. set 1)

“function sets” refer to groups of functions that together can be used to solve a given question

Dataset Creation

Function wrapping

Func A: Get the ID and salary of employees who report to a manager with a specific first name

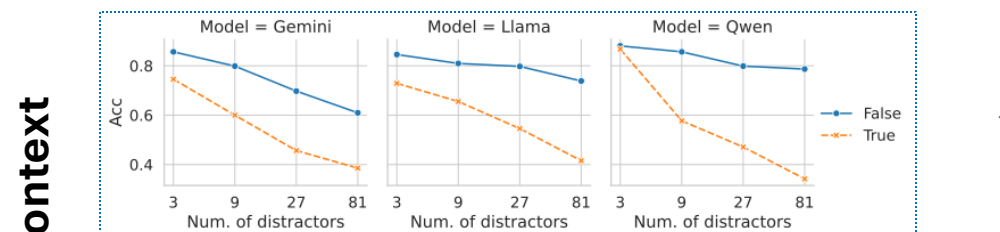
Func B: Get the ID for a specific first name

Func C: Get ID and salary for employees under a manager with a specific ID

Function set 1

Function set 2

Tool schemas provided in prompt



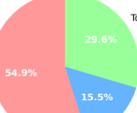
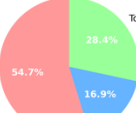
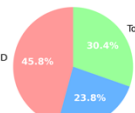
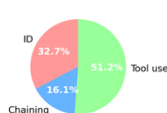
In-context

3 Tools
Total failures: 211

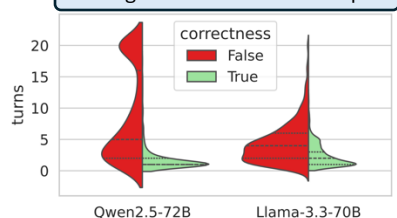
9 Tools
Total failures: 332

27 Tools
Total failures: 451

81 Tools
Total failures: 510



Taking more turns doesn't help.



The share of tool identification errors increases with the number of tool schemas in the prompt

RAG with a vector DB of 4450 tool schemas

Model	External errors (N)	External errors (Y)	Acc. decrease
Gemini 2.0 (Flash)	71.4 ± 1.6	41.1 ± 1.7	42.4%
GPT 4o	60.5 ± 1.7	38.4 ± 1.7	36.5%
Llama 3.3 (70b, Instruct)	64.0 ± 1.7	38.9 ± 1.7	39.2%
Llama 3.1 (70b, Instruct)	42.3 ± 1.7	23.3 ± 1.5	44.9%
Qwen 2.5 (72B, Instruct)	64.1 ± 1.6	35.3 ± 1.7	44.9%

RAG

Model	Total failures	Search	ID	Chaining	Tool use
Gemini 2.0 (Flash)	489	57.1	25.5	12.3	6.14
GPT 4o	511	52.6	24.6	8.61	14.1
Llama 3.3 (70b, Instruct)	496	66.3	11.9	13.9	7.86
Llama 3.1 (70b, Instruct)	495	59.3	18.0	11.7	10.9
Qwen 2.5 (72B, Instruct)	637	60.5	26.4	7.06	6.12

What's Next?

- Agent memory and test-time learning:** Can agents remember which tools to avoid? Can memory modules improve performance?
- Inference-time compute:** Does thinking more help agents identify backup plans?



Paper



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