

Grokking the Agents

How LLM Coding Assistants Actually Work Under the Hood — A Case Study with a Simple Example

Oliver Dürr, Sep 2025

This is a brown bag seminar!

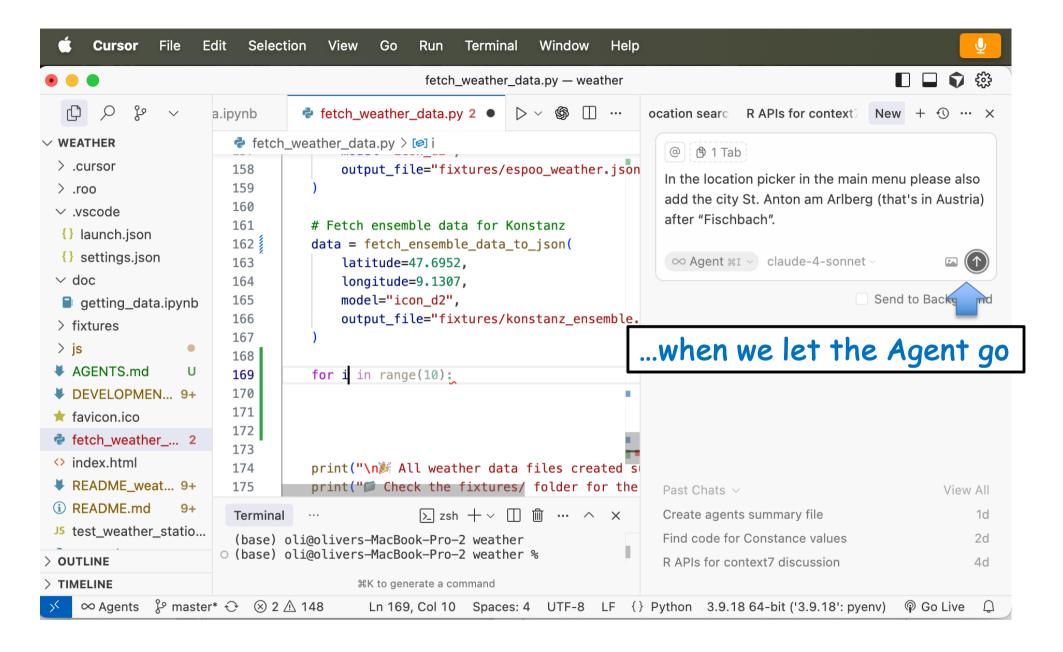
- Wanted: Talks zu Methoden, Papers, Erfahrungen auf Konferenzen, Ideen, ...
- Perfekt gestylten Power-Points sind nicht nötig, Tafel tut's auch
- Die Vortragenden müssen keine Experten sein, dumme Fragen sind erwünscht
- Keine Marketing-Veranstaltung

Die Vortragenden müssen keine Experten sein, dumme Fragen sind erwünscht.

Disclaimer: I look at Agentic Coding but do not consider myself as an expert in that field.

Introduction

What happens...

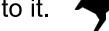


Coding Agents Selection (Sep, 2025)

Agent / Tool	Open Source	CLI Tool	IDE Plugin	Full IDE
Cursor	X	X	X	V
Windsurf	X	X	X	V
Codex (OpenAI, 2025)	V	V	V	X
Claude Code	X	V	V	X
Copilot	X	X	V	X
Roo Code	V	V	\checkmark	X
Kilo Code	V	V	V	X

- Cursor / Windsurf extra IDE for AI-based coding
- Plugins are inside IDEs (typical VS-Code / Cursor)
- CLI command line tool started w/o IDE





LLMs The brains we are using

Step 0 Tokenization Text → Numbers (Tokens)

 A simple component takes text and translates that into integer numbers

The weather is really nice today. I'm thinking about going for a

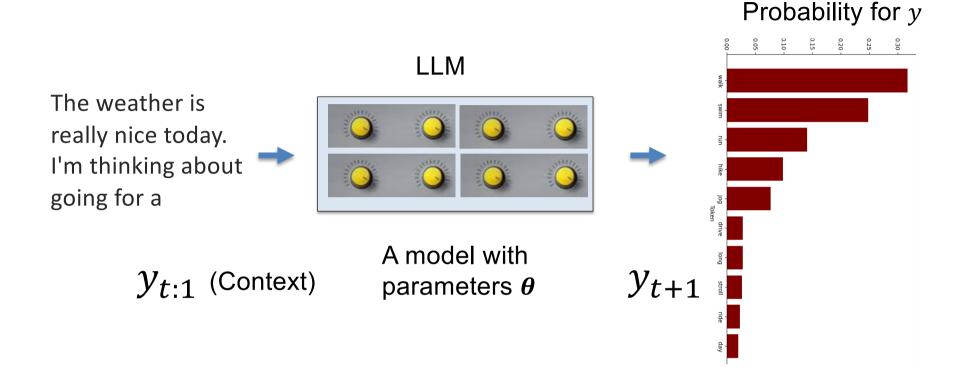
Tokens Characters

14 66

The weather is really nice today. I'm thinking about going for a

 $y_{14:1}$ =[976, 11122, 382, 2715, 7403, 4044, 13, 5477, 9791, 1078, 2966, 395, 261, 793]

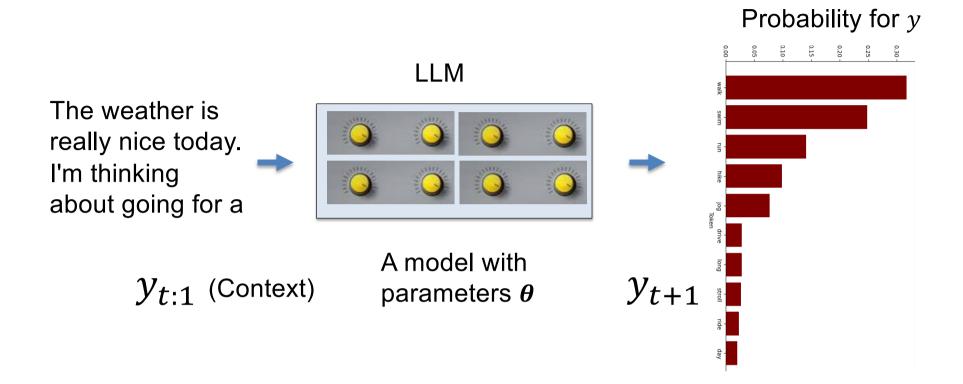
LLMs are probabilistic models



In math an LLM is

$$p_{\boldsymbol{\theta}}(y_{t+1}|y_{t:1})$$

LLMs are probabilistic models



In code an LLM is

```
# Input: sequence of tokens y[0:(t-1)] and weights θ
# Output: probability distribution over next token y[t]
function next_token_distribution(y[0:t], θ):
    ...
    return probs # = pθ(y_{t} | y_{0:(t-1)})
```

Creating response in an autoregressive fashion



Describe your technology in one sentence.



I am a Generative Language model based on the transformer architecture that uses self-attention mechanisms and is trained with maximum likelihood principle to predict the next token in a sequence of tokens and generate new text.





At each step, the model samples the next token from $p(y_{t+1}|y_{t:1})$ or takes most likely token (when temperature T=0)

Steps 1: Describe your technology in one sentence. \rightarrow I

Steps 2: Describe your technology in one sentence. $I \rightarrow am$

Steps 3: Describe your technology in one sentence. I am \rightarrow a

Steps 4: Describe your technology in one sentence. I am a \rightarrow generative

Step 36 Describe your technology in one sentence. I am a ... new text. → END

Important facts about LLM $p_{\theta}(y_{t+1}|y_{t:0})$

Stateless Models

- Every call is independent $p_{\theta}(y_{t+1}|y_{t:0})$ has no memory
- Native LLM don't remember last conversation.

Limited Context ("window size")

• Each model can only process a finite number of tokens $t \leq T_{max}^*$

This requires very tricky engineering (context engineering)

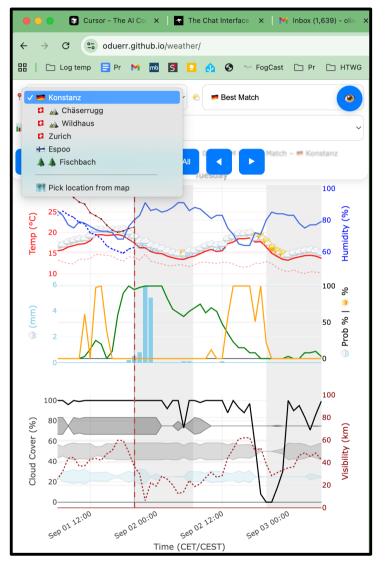
Examples Context Sizes of Current LLMs

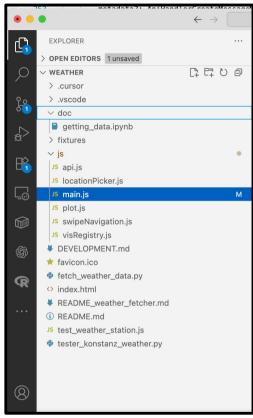
- GPT-5 → 400K tokens
- Grok Code Fast → 256K tokens
- Claude Sonnet 4 → 200K (Version with 1M)
- Qwen 3 → 128K tokens (self-hostable)
- DeepSeek-Coder V2 → 128K tokens (self-hostable)

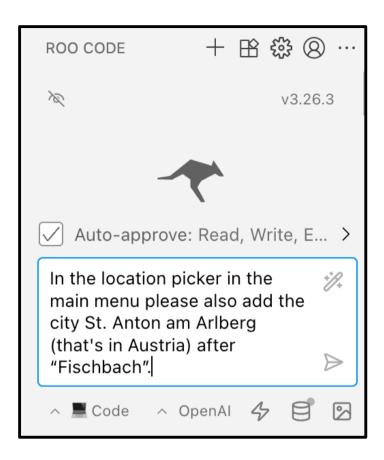
13

Case Study

Very Simple Task in existing code base



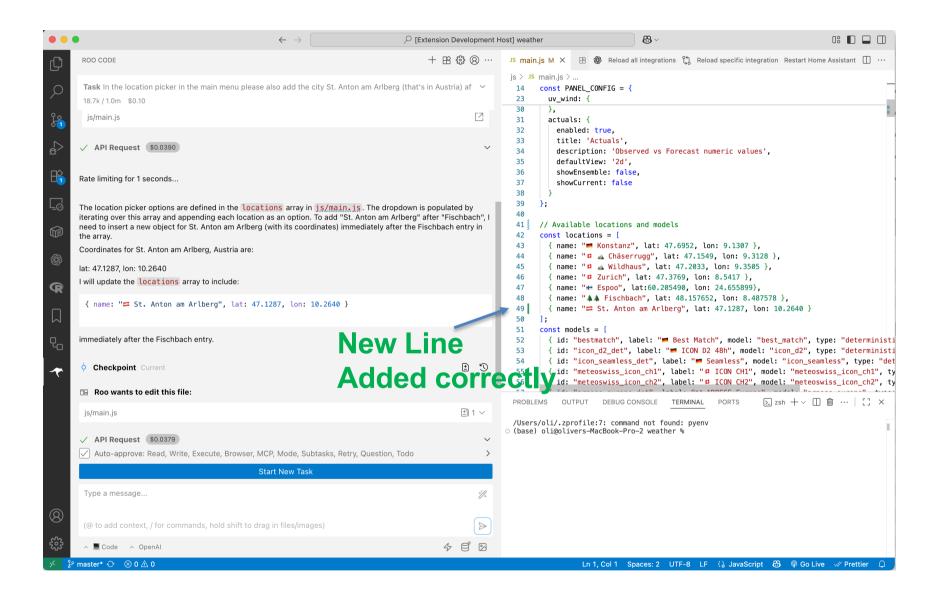




Mini refactoring of a small, partly vibe-coded project. Showing results of weather models for certain locations.

<u>https://oduerr.github.io/weather/</u> (gh tag is demo-baseline)

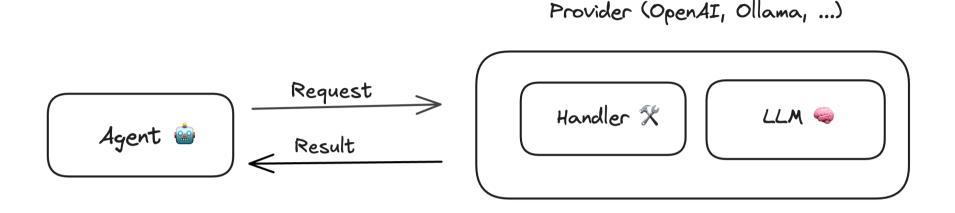
Result, using Roo Code and gpt-4.1



Only one line of code has been changed, perfect!

How does this work (highlevel)

How Agents Communicate with LLM Providers (one shot)



- The Agent (Roo/Cursor) sends a request to the Provider (e.g. OpenAl, Ollama, ...).
- The Handler X makes a single string from the request (e.g. system prompt / user prompt*), tokenizes it to $y_{t:1}$ and sends it to the LLM.
- The LLM $\stackrel{\text{\tiny ω}}{=}$ calculates the probability distribution $p(y_{t+1}|y_{t:1})$ of next token y_{t+1}
- The Handler repeatably samples the tokens and sends them it to the agent

^{*}Also, might add context from previous conversations.

Creating the prompt (Conceptually)

- The system prompt and the user prompt (optionally other prompts) are sent to the provider
- A single prompt / string is then built using a template like

```
<|system|>
You are Roo, a helpful...
<|user|>
Please add the location St. Anton...
```

• Important: What enters LLM is just text / tokens $y_{t:1}$

 The handler then repeatedly samples tokens until the response is complete.

Under the Hood

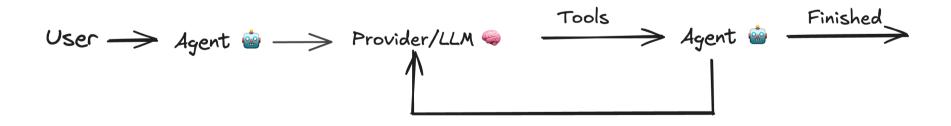
- Providers use optimizations and tricks beyond simple prompt assembly
 - KV-cache reuse
 - Caching the internals of the transformer which take advantage of the autoregressive nature
 - Prompt caching
 - System prompt (and other input at the beginning stays constant)
 - Transformer architecture allows for caching the beginning of network

Likely providers such as OpenAl also have a secret sauce.

The agentic loop

- This usually happens in a loop (aka <u>Re</u>ason <u>Act</u> Loop, ReAct)
 - 1. The Agent sends the system prompt ("You are Roo, a helpful...") and the user prompt ("Please add the location St. Anton...") to the Provider.
- 2. The Provider returns the response of the LLM to the Agent (usually request to use a tool).
- 3. The Agent fulfills the request (e.g. calls the tool) and sends the result back to the LLM.

Step 2 and 3 are iterated until success (kind of while loop)*

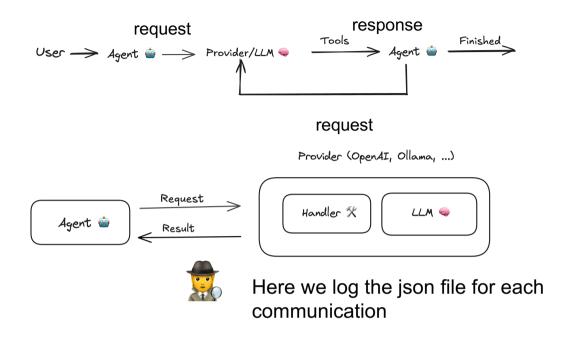


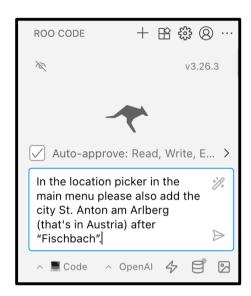
ReAct: https://arxiv.org/abs/2210.03629

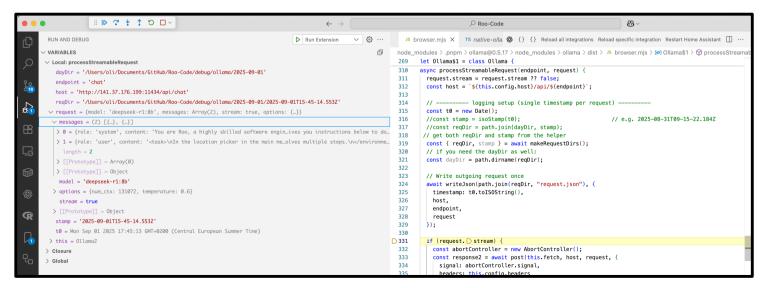
^{*}In principle, also other non-loop-like calls are possible.

Eavesdropping the communication

Eavesdropping communication







Three rounds happend

Using openAl's GPT-4.1, three rounds in the loop happened (see json files)

openai-log-viewer.html	Yesterday at 22:41	35 KB HTML text
→ im Ilm_input	Yesterday at 21:50	Folder
openai-native-input-2025-09-01T17-56-34-674Z.json	Yesterday at 19:56	44 KB JSON File
openai-native-input-2025-09-01T17-55-33-852Z.json	Yesterday at 19:55	76 KB JSON File
openai-native-input-2025-09-01T17-54-33-066Z.json	Yesterday at 19:54	44 KB JSON File
✓ 🛅 llm_output	Yesterday at 21:50	Folder
openai-native-output-2025-09-01T17-56-43-623Z.json	Yesterday at 19:56	874 bytes JSON File
openai-native-output-2025-09-01T17-55-42-902Z.json	Yesterday at 19:55	2 KB JSON File
openai-native-output-2025-09-01T17-54-39-349Z.json	Yesterday at 19:54	1 KB JSON File

- openai-log-viewer.html is a vibe-coded simple viewer for the json files
- Files are available at
- https://www.dropbox.com/scl/fo/8gtzr2qi3boil0r8fhiqy/AKdfu4WPl7M9PXIufd348M?rlkey=z46p92u1ps2tao7rtarg7wd0x&st=s3zjd nwd&dl=0

Step 1 Agent to provider/LLM (Principle) > > > Raw Request





```
44167 characters
"requestBody": {
  "model": "gpt-4.1",
                                              User msg
      "role": "user".
      "content": [
          "type": "input_text",
          "text": "<task>\nIn the location picker in the main menu please also add the city St. Anton am Arlberg (that's in Austria) after
          "type": "input_text",
          "text": "<environment_details>\n# VSCode Visible Files\njs/main.js\n\n# VSCode Open Tabs\njs/main.js\n\n# Current Time\nCurrent
  "stream": true,
  "store": true,
  "instructions": "You are Roo, a highly skilled software engineer with extensive knowledge in many programming languages, frameworks, des
  "temperature": 0,
  "max_output_tokens": 32768
},
                                             System msg
"model": {
  "id": "gpt-4.1",
    "maxTokens": 32768,
    "contextWindow": 1047576,
    "supportsImages": true,
    "supportsPromptCache": true,
    "inputPrice": 2,
    "outputPrice": 8,
    "cacheReadsPrice": 0.5,
    "supportsTemperature": true
  "format": "openai",
  "maxTokens": 32768,
  "temperature": 0
"metadata": {
 "mode": "code",
  "taskId": "43f2c6e9-cfd9-4813-b90b-d9ef7329d341"
```

Round 1 System msg 41415 chars, first entries

You are Roo, a highly skilled software engineer with extensive knowledge in many programming languages, frameworks, design patterns, and best practices.

==== MARKDOWN RULES

• • •

TOOL USE

You have access to a set of tools that are executed upon the user's approval. You can use one tool per message, and will receive the result of that tool use in the user's response. You use tools step-by-step to accomplish a given task, with each tool use informed by the result of the previous tool use.

Round 1 System Msg cont'd (Tools)

```
Explaining the tool read file for the LLM
# Tools
## read file
Description: Request to read the contents of one or more files. The tool outputs line-numbered
content (e.g. "1 | const x = 1") for easy reference when creating diffs or discussing code. Supports text extraction from PDF and DOCX files, but may not handle other binary files
properly.
**IMPORTANT: You can read a maximum of 5 files in a single request.** If you need to read more
files, use multiple sequential read_file requests.
Parameters:
- args: Contains one or more file elements, where each file contains:
  - path: (required) File path (relative to workspace directory
/Users/oli/Documents/GitHub/weather)
Usaae:
<read file>
<aras>
  <file>
    <path>path/to/file</path>
  </file>
</aras>
</read file>
```

Round 1 System Msg cont'd Addition Tools (in system prompt, GPT-summarized)

From the JSON you showed (the system prompt's **Tools** section), the LLM is allowed to call these tools:

```
read file — read up to 5 files (with line numbers).
fetch instructions — fetch "how-to" instructions for tasks (e.g., create mcp server, create mode).
search files — regex search across a directory, returns matches with context.
list files — list directory contents (optionally recursive).
list code definition names — extract function/class/method names from files/dirs.
apply diff — make surgical search/replace edits in a file via precise diff blocks.
write to file — create/overwrite a file with full content (single-shot).
insert content — insert new lines into a file at a given line (or append).
search and replace — find/replace (literal or regex) within a file, with optional line range.
execute command — run a CLI command (with optional working directory).
ask followup question — ask the user for missing details (with suggested answers).
attempt completion — finalize the task and present the result (only after confirming prior tool
success).
switch mode — switch to another mode (code / ask / architect / debug / orchestrator).
new task — start a new task instance in a chosen mode.
update todo list — replace the whole TODO checklist with updated statuses.
```

The green one have been picked up be the agent.

Round 1 User Msg

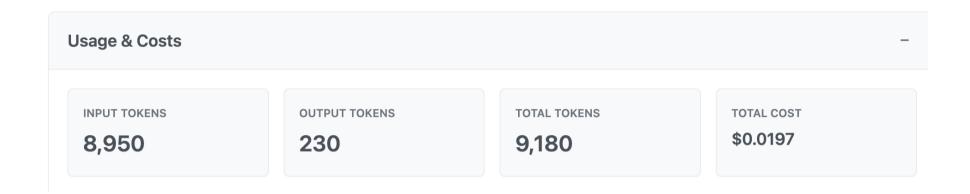
Text that I entered

```
<task>
In the location picker in the main menu please also add the city St. Anton am Arlberg (that's in Austria) after "Fischbach".
</task>
<environment details>
# VSCode Visible Files
js/main.js
# VSCode Open Tabs
js/main.js
# Current Time
Current time in ISO 8601 UTC format: 2025-09-01T17:54:33.033Z
User time zone: Europe/Berlin, UTC+2:00
# Current Workspace Directory (/Users/oli/Documents/GitHub/weather) Files
DEVELOPMENT.md
favicon.ico
fetch_weather_data.py
index.html
README_weather_fetcher.md
README.md
test_weather_station.js
tester_konstanz_weather.py
.cursor/
.vscode/
doc/
doc/getting_data.ipynb
</environment details>
```

Round 1 Response from LLM (GPT-4.1)

```
<thinking>
The user wants to add "St. Anton am Arlberg" (Austria) to the location
picker in the main menu, specifically after "Fischbach". Based on the file structure, the location picker logic is likely in ['js/locationPicker.js'](js/locationPicker.js:1), but it could also be referenced or rendered in ['js/main.js'](js/main.js:1). Since the user has both files available and the picker is likely defined or at least configured in 'locationPicker.js', I should read both files to find where
the list of locations is defined and how to insert the new city after
"Fischbach".
I will read both [`js/locationPicker.js`](js/locationPicker.js:1) and
[`js/main.js`](js/main.js:1) to get the full context and ensure I update
the correct place.
</thinking>
<read file>
<arqs>
                                                                                             Tool call for the two files
    <file>
        <path>js/locationPicker.js</path>
    </file>
    <file>
        <path>js/main.js</path>
    </file>
</arqs>
</read file>
```

Summary of round 1



Preparation for round 2

The agent now uses the tool / executes the command. Here it reads the 2 files (js/locationPicker.js) and (js/main.js) and appends them to the user msg

Round 2 (raw reguest)

```
76308 characters
"requestBody": {
 "model": "gpt-4.1",
     "role": "user",
                                                       User msg (new)
     "content": [
         "type": "input_text",
         "text": "[read_file for 'js/locationPicker.js', 'js/main.js'] Result:"
         "type": "input_text",
         "text": "<file>>\n<file><path>js/locationPicker.js</path>\n<content lines=\"1-323\">\n 1 | /**\n 2 | * Location Picker Module
         "type": "input_text",
         "text": "<environment_details>\n# VSCode Visible Files\nis/main.is\n\n# VSCode Open Tabs\nis/main.is\n\n# Current Time\nCurrent
  "stream": true,
 "store": true,
 "instructions": "You are Roo, a highly skilled software engineer with extensive knowledge in many programming languages, frameworks, des
 "temperature": 0,
 "max_output_tokens": 32768,
 "previous_response_id": "resp_68b5ddd9\\d4819193fc38d2d3b94d1f027f2fdceca5f9b3"
"model": {
 "id": "gpt-4.1",
                                                     System msg, stays the same
   "maxTokens": 32768,
   "contextWindow": 1047576,
   "supportsImages": true,
   "supportsPromptCache": true,
   "inputPrice": 2,
   "outputPrice": 8,
   "cacheReadsPrice": 0.5,
   "supportsTemperature": true
  "format": "openai",
 "maxTokens": 32768,
 "temperature": 0
"metadata": {
 "mode": "code",
 "taskId": "43f2c6e9-cfd9-4813-b90b-d9ef7329d341"
```

Round 2 (User prompt, with additional file information)

```
[read_file for 'js/locationPicker.js', 'js/main.js'] Result:
<files>
<file><path>js/locationPicker.js</path>
                                                               Content of the first
<content lines="1-323">
  1 | /**
                                                               file.
       * Location Picker Module
  3 | * Provides map-based location selection functionality
  4
<file><path>js/main.js</path>
<content lines="1-482">
  1 | let measured_temp = null;
  2 | let measured_water_temp = null;
                                                               Content of the second
                                                               file.
41 | // Available locations and models
 42 | const locations = \Gamma
 43 I
        { name: "■ Konstanz", lat: 47.6952, lon: 9.1307 },
        { name: "□ 🚵 Chäserrugg", lat: 47.1549, lon: 9.3128 },
 44
      { name: "♬ 🔌 Wildhaus", lat: 47.2033, lon: 9.3505 },
 45 I
 46 | { name: "□ Zurich", lat: 47.3769, lon: 8.5417 },
       { name: "
    Espoo", lat:60.205490, lon: 24.655899},
47 I
       { name: "* Fischbach", lat: 48.157652, lon: 8.487578 }
 48 I
 49 | ];
```

Round 2: Response

<thinking>

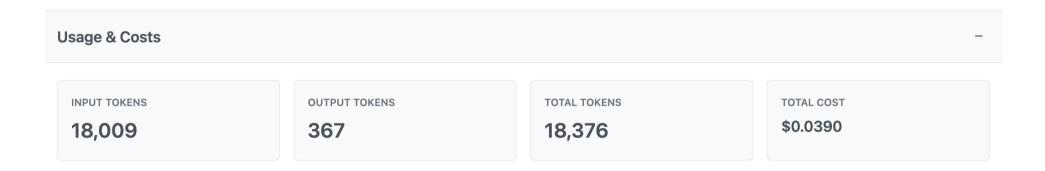
The location picker options are defined in the `locations` array in [`js/main.js`](js/main.js:42-49). The dropdown is populated by iterating over this array and appending each location as an option. To add "St. Anton am Arlberg" after "Fischbach", I need to insert a new object for St. Anton am Arlberg (with its coordinates) immediately after the Fischbach entry in the array.

```
Coordinates for St. Anton am Arlberg, Austria are:
lat: 47.1287, lon: 10.2640
I will update the `locations` array to include:
```is
{ name: "■ St. Anton am Arlberg", lat: 47.1287, lon: 10.2640 }
immediately after the Fischbach entry.
</thinkina>
<apply_diff>
<path>js/main.js</path>
<diff>
<<<<< SEARCH
:start_line:47
 { name: "≝ Espoo", lat:60.205490, lon: 24.655899},
 { name: "* Fischbach", lat: 48.157652, lon: 8.487578 }
 { name: "* Fischbach", lat: 48.157652, lon: 8.487578 },
 name: "St. Anton am Arlberg", lat: 47.1287, lon: 10.2640
>>>>> REPLACE
</diff>
</apply_diff>
```

It knows the location of Arlberg, out of the box. This needs checking!

That's a tool call

# Round 2: Summary



The diff is executed and send to the provider/LLM

### Round 3 (Finishing off)

#### User (System stays like it is)

```
[apply_diff for 'js/main.js'] Result:

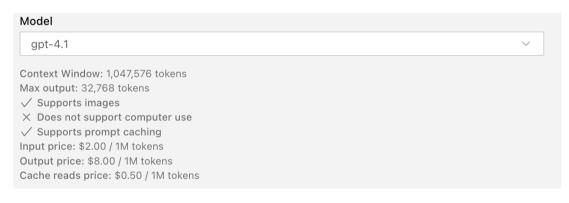
<file_write_result>
<path>js/main.js</path>
<operation>modified</operation>
<notice>
<i>You do not need to re-read the file, as you have seen all changes</i>
<i>Proceed with the task using these changes as the new baseline.</i>
</notice>
</file_write_result>
```

<attempt\_completion>
<result>
"St. Anton am Arlberg" (Austria) has been added to the location picker immediately after "Fischbach" in the main menu. The new entry uses the correct coordinates and will now appear in the dropdown for selection.
</result>
</attempt\_completion>

#### Costs

Costs For the single line change





GPT5 is (currently) cheaper! Also codex-mini-latest \$0.03

#### Sidenote Be skeptical and don't trust the LLMs!

The line added was
{ name: "St. Anton am Arlberg",
lat: 47.1287, lon: 10.2640 }

Generate google maps link for lat: 47.1287, lon: 10.2640

 This time it worked but I also had runs with hallucinations. https://www.google.com/maps/search/?api=1&query=47.1287%2C10.2640



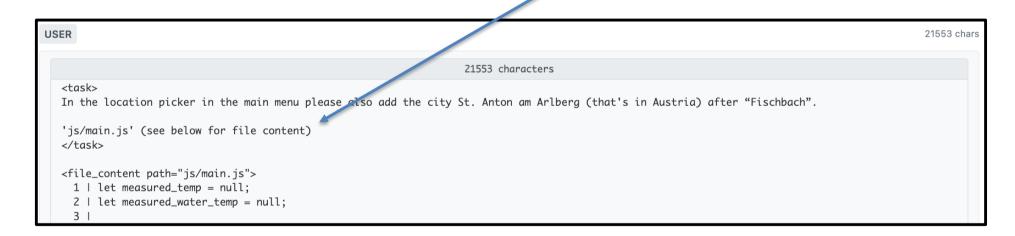
# Practical takeaways

# Adding Content (in the prompt)

You can help the Agent by providing extra content

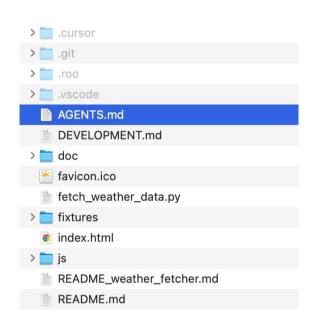


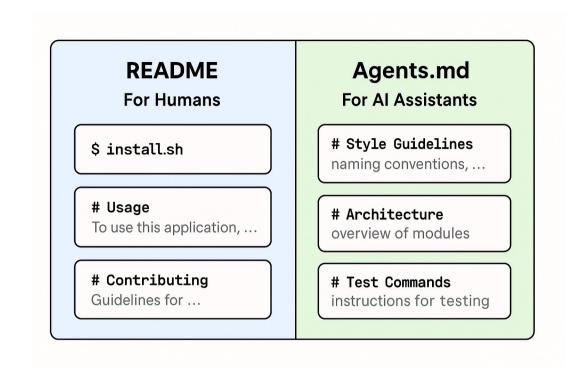
Add the content of js/main.js to the user prompt



GPT4.1 then does the job in a single round (Ollma models still fail).

### Project Context (AGENTS.md)





If present the file AGENTS.md is added always after the system prompt. All major agents: cursor, codex, Roo supports this now (August 2025). Replaces custom solutions (like CLAUDE.md)

See <a href="https://agents.md/">https://agents.md/</a>

# Example for AGENTS.MD for Deep Learning

```
Tell it to use Keras
AGENTS.md - Keras-first, PyTorch fallback
Policy
- Implement everything in **Keras 3** by default.
- Fallback to **PyTorch** if Keras cannot express the feature or if the user
requests it.
- All tests use the **pytest** framework.
 Tell it which tests to run
Protocol
1. Apply minimal diffs
2. Validate with **unit + smoke tests**
 - You may skip the **long test suite**.
Device
- Auto-select: `cuda` if available → else `mps` (Apple) → else `cpu`.
Tests layout
tests/
├ unit/
 # fast checks
\vdash smoke/ # tiny training runs (\leq2 batches, CPU OK)
└ long/
 # integration/benchmark (may skip)
```

#### Context Management in Cursor

#### Use Cursor Rules AGENTS.md

- Define project-wide behaviors like "always use TypeScript import"
- These are always included in the context

#### Add Context in Prompt

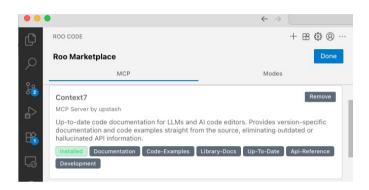
You can webpage with documentation and best practices (use @)

#### Context Management by Cursor

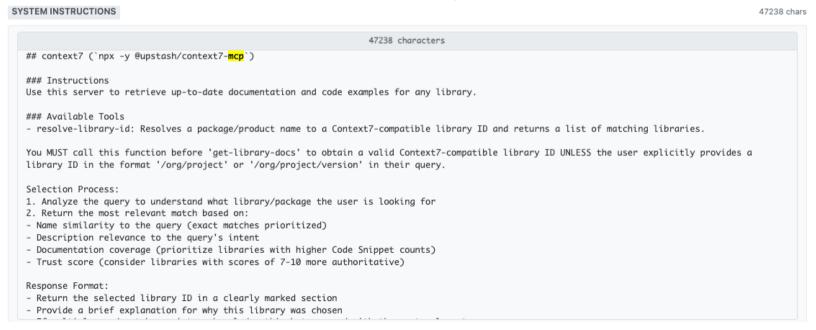
- Automatic context building: Cursor includes files you open or edit into the LLM context.
- Smart truncation: If files are too large, only the most relevant parts are sent.
- Session memory: Keeps track of recent edits & conversations, discards irrelevant history uses RAG (cursor's secret sauce)

#### Adding Tools via MCP Protocol (Roo-Code)

You can add your own tools using the MCP protocol (here context7)



The added MCP Server now appears in the System Instruction.

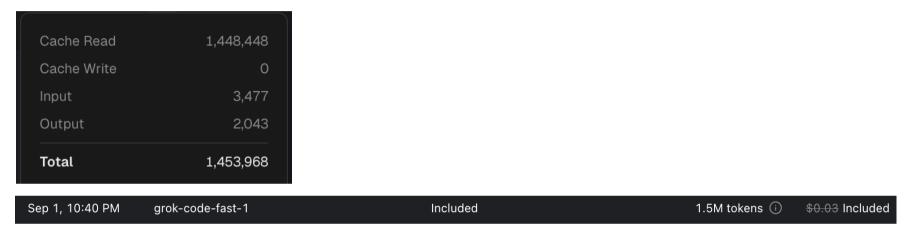


#### Special MCP Servers for Coding

- Context7 has documentation of 15k libraries (including R's ggplot :-)
  - For Cursor install via <a href="https://docs.cursor.com/en/tools/mcp">https://docs.cursor.com/en/tools/mcp</a>
  - Use something like "use context7" in prompt
- Other MCP Servers
  - Github
  - MCP servers for Memory Management
    - Better than the ones provided by the Agents?

### Costs / Locally Hosted LLM

 For the adding Arlberg task grok-code-fast-1 needed approx 12 seconds (using cursor), also running linter test.



Costs for my extensive coding streak (2 weeks) sum up to ~ 100 USD

- Local LLM could not do the task?
  - only tried 8B models (probably the wrong ones) and did not try hard
  - My belief, just the matter of time when small specialized local LLMs can better use the tools the agents provide them.

#### How good is it?

- https://www.swebench.com/
  - Task: Create fixes in GitHub repositories
  - Different versions of the benchmark also include a standard agent





From 11% to 60% success in 2 years

### Literature / Links (fast moving field)

- SWE-Bench (Benchmarks and simple agent in Python\*!)
  - https://www.swebench.com/original.html
  - https://www.youtube.com/watch?v=yAQw77juYSg

#### A lot of hype noise

- Reddit: <a href="https://reddit.com/r/cursor">https://reddit.com/r/cursor</a> r/AI\_Agents/ r/ChatGPTCoding
- Youtube
  - Andrej Karpathy: Software Is Changing (Opinion)
    - https://youtu.be/LCEmiRjPEtQ?si=YDpMXjbRqZDKVZfS
  - https://www.youtube.com/@matthew\_berman (Channel)

#### Love to get your input!

Also, on applications beyond coding (we plan to use it in the context of Statistics / Causality)