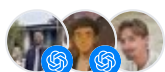



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# GPT-5 prompting guide



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GPT-5, our newest flagship model, represents a substantial leap forward in agentic task performance, coding, raw intelligence, and steerability.

While we trust it will perform excellently “out of the box” across a wide range of domains, in this guide we’ll cover prompting tips to maximize the quality of model outputs, derived from our experience training and applying the model to real-world tasks. We discuss concepts like improving agentic task performance, ensuring instruction adherence, making use of newly API features, and optimizing coding for frontend and software engineering tasks - with key insights into AI code editor Cursor’s prompt tuning work with GPT-5.

We’ve seen significant gains from applying these best practices and adopting our canonical tools whenever possible, and we hope that this guide, along with the [prompt optimizer tool](#) we’ve built, will serve as a launchpad for your use of GPT-5. But, as always, remember that prompting is not a one-size-fits-all exercise - we encourage you to run experiments and iterate on the foundation offered here to find the best solution for your problem.

## Agentic workflow predictability

We trained GPT-5 with developers in mind: we’ve focused on improving tool calling, instruction following, and long-context understanding to serve

as the best foundation model for agentic applications. If adopting GPT-5 for agentic and tool calling flows, we recommend upgrading to the [Responses API](#), where reasoning is persisted between tool calls, leading to more efficient and intelligent outputs.

## Controlling agentic eagerness

Agentic scaffolds can span a wide spectrum of control—some systems delegate the vast majority of decision-making to the underlying model, while others keep the model on a tight leash with heavy programmatic logical branching. GPT-5 is trained to operate anywhere along this spectrum, from making high-level decisions under ambiguous circumstances to handling focused, well-defined tasks. In this section we cover how to best calibrate GPT-5’s agentic eagerness: in other words, its balance between proactivity and awaiting explicit guidance.

### Prompting for less eagerness

GPT-5 is, by default, thorough and comprehensive when trying to gather context in an agentic environment to ensure it will produce a correct answer. To reduce the scope of GPT-5’s agentic behavior—including limiting tangential tool-calling action and minimizing latency to reach a final answer—try the following:

- Switch to a lower `reasoning_effort` . This reduces exploration depth but improves efficiency and latency. Many workflows can be accomplished with consistent results at medium or even low `reasoning_effort` .
- Define clear criteria in your prompt for how you want the model to explore the problem space. This reduces the model’s need to explore and reason about too many ideas:

```
<context_gathering>
```

```
Goal: Get enough context fast. Parallelize discovery and stop as soon as you ca
```

```
Method:
```

- Start broad, then fan out to focused subqueries.
- In parallel, launch varied queries; read top hits per query. Deduplicate path
- Avoid over searching for context. If needed, run targeted searches in one par

```
Early stop criteria:
```

- You can name exact content to change.
- Top hits converge (~70%) on one area/path.

```
Escalate once:
- If signals conflict or scope is fuzzy, run one refined parallel batch, then p
Depth:
- Trace only symbols you'll modify or whose contracts you rely on; avoid transi
Loop:
- Batch search → minimal plan → complete task.
- Search again only if validation fails or new unknowns appear. Prefer acting o
</context_gathering>
```

If you're willing to be maximally prescriptive, you can even set fixed tool call budgets, like the one below. The budget can naturally vary based on your desired search depth.

```
<context_gathering>
- Search depth: very low
- Bias strongly towards providing a correct answer as quickly as possible, even
- Usually, this means an absolute maximum of 2 tool calls.
- If you think that you need more time to investigate, update the user with you
</context_gathering>
```

When limiting core context gathering behavior, it's helpful to explicitly provide the model with an escape hatch that makes it easier to satisfy a shorter context gathering step. Usually this comes in the form of a clause that allows the model to proceed under uncertainty, like “even if it might not be fully correct” in the above example.

### Prompting for more eagerness

On the other hand, if you'd like to encourage model autonomy, increase tool-calling persistence, and reduce occurrences of clarifying questions or otherwise handing back to the user, we recommend increasing `reasoning_effort`, and using a prompt like the following to encourage persistence and thorough task completion:

```
<persistence>
- You are an agent - please keep going until the user's query is completely res
- Only terminate your turn when you are sure that the problem is solved.
- Never stop or hand back to the user when you encounter uncertainty – research
- Do not ask the human to confirm or clarify assumptions, as you can always ad
</persistence>
```

Generally, it can be helpful to clearly state the stop conditions of the agentic tasks, outline safe versus unsafe actions, and define when, if ever, it's acceptable for the model to hand back to the user. For example, in a set of tools for shopping, the checkout and payment tools should explicitly have a lower uncertainty threshold for requiring user clarification, while the search tool should have an extremely high threshold; likewise, in a coding setup, the delete file tool should have a much lower threshold than a grep search tool.

## Tool preambles

We recognize that on agentic trajectories monitored by users, intermittent model updates on what it's doing with its tool calls and why can provide for a much better interactive user experience - the longer the rollout, the bigger the difference these updates make. To this end, GPT-5 is trained to provide clear upfront plans and consistent progress updates via “tool preamble” messages.

You can steer the frequency, style, and content of tool preambles in your prompt—from detailed explanations of every single tool call to a brief upfront plan and everything in between. This is an example of a high-quality preamble prompt:

```
<tool_preambles>
- Always begin by rephrasing the user's goal in a friendly, clear, and concise
- Then, immediately outline a structured plan detailing each logical step you'll
- Finish by summarizing completed work distinctly from your upfront plan.
</tool_preambles>
```

Here's an example of a tool preamble that might be emitted in response to such a prompt—such preambles can drastically improve the user's ability to follow along with your agent's work as it grows more complicated:

```
"output": [
  {
    "id": "rs_6888f6d0606c819aa8205ecce386963f0e683233d39188e7",
    "type": "reasoning",
    "summary": [
      {
```

```

        "type": "summary_text",
        "text": "***Determining weather response**\n\nI need to answer the use
    },
},
{
    "id": "msg_6888f6d83acc819a978b51e772f0a5f40e683233d39188e7",
    "type": "message",
    "status": "completed",
    "content": [
        {
            "type": "output_text",
            "text": "I\u2019m going to check a live weather service to get the cu
        }
    ],
    "role": "assistant"
},
{
    "id": "fc_6888f6d86e28819aaaa1ba69cca766b70e683233d39188e7",
    "type": "function_call",
    "status": "completed",
    "arguments": "{\"location\": \"San Francisco, CA\", \"unit\": \"f\"}",
    "call_id": "call_XOnF4B9DvB8EJVB3JvWnGg83",
    "name": "get_weather"
},
],

```

## Reasoning effort

We provide a `reasoning_effort` parameter to control how hard the model thinks and how willingly it calls tools; the default is `medium`, but you should scale up or down depending on the difficulty of your task. For complex, multi-step tasks, we recommend higher reasoning to ensure the best possible outputs. Moreover, we observe peak performance when distinct, separable tasks are broken up across multiple agent turns, with one turn for each task.

## Reusing reasoning context with the Responses API

We strongly recommend using the Responses API when using GPT-5 to unlock improved agentic flows, lower costs, and more efficient token usage in your applications.

We've seen statistically significant improvements in evaluations when

using the Responses API over Chat Completions—for example, we observed Tau-Bench Retail score increases from 73.9% to 78.2% just by switching to the Responses API and including `previous_response_id` to pass back previous reasoning items into subsequent requests. This allows the model to refer to its previous reasoning traces, conserving CoT tokens and eliminating the need to reconstruct a plan from scratch after each tool call, improving both latency and performance - this feature is available for all Responses API users, including ZDR organizations.

## Maximizing coding performance, from planning to execution

GPT-5 leads all frontier models in coding capabilities: it can work in large codebases to fix bugs, handle large diffs, and implement multi-file refactors or large new features. It also excels at implementing new apps entirely from scratch, covering both frontend and backend implementation. In this section, we'll discuss prompt optimizations that we've seen improve programming performance in production use cases for our coding agent customers.

### Frontend app development

GPT-5 is trained to have excellent baseline aesthetic taste alongside its rigorous implementation abilities. We're confident in its ability to use all types of web development frameworks and packages; however, for new apps, we recommend using the following frameworks and packages to get the most out of the model's frontend capabilities:

- Frameworks: Next.js (TypeScript), React, HTML
- Styling / UI: Tailwind CSS, shadcn/ui, Radix Themes
- Icons: Material Symbols, Heroicons, Lucide
- Animation: Motion
- Fonts: San Serif, Inter, Geist, Mona Sans, IBM Plex Sans, Manrope

### Zero-to-one app generation

GPT-5 is excellent at building applications in one shot. In early

experimentation with the model, users have found that prompts like the one below—asking the model to iteratively execute against self-constructed excellence rubrics—improve output quality by using GPT-5’s thorough planning and self-reflection capabilities.

```
<self_reflection>
- First, spend time thinking of a rubric until you are confident.
- Then, think deeply about every aspect of what makes for a world-class one-shot
- Finally, use the rubric to internally think and iterate on the best possible
</self_reflection>
```

## Matching codebase design standards

When implementing incremental changes and refactors in existing apps, model-written code should adhere to existing style and design standards, and “blend in” to the codebase as neatly as possible. Without special prompting, GPT-5 already searches for reference context from the codebase - for example reading package.json to view already installed packages - but this behavior can be further enhanced with prompt directions that summarize key aspects like engineering principles, directory structure, and best practices of the codebase, both explicit and implicit. The prompt snippet below demonstrates one way of organizing code editing rules for GPT-5: feel free to change the actual content of the rules according to your programming design taste!

```
<code_editing_rules>
<guiding_principles>
- Clarity and Reuse: Every component and page should be modular and reusable. A
- Consistency: The user interface must adhere to a consistent design system—col
- Simplicity: Favor small, focused components and avoid unnecessary complexity
- Demo-Oriented: The structure should allow for quick prototyping, showcasing t
- Visual Quality: Follow the high visual quality bar as outlined in OSS guideli
</guiding_principles>
<frontend_stack_defaults>
- Framework: Next.js (TypeScript)
- Styling: TailwindCSS
- UI Components: shadcn/ui
- Icons: Lucide
- State Management: Zustand
- Directory Structure:
  \`\`\`
/src
/app
```

```
    /api/<route>/route.ts      # API endpoints
    /(pages)                   # Page routes
    /components/               # UI building blocks
    /hooks/                    # Reusable React hooks
    /lib/                      # Utilities (fetchers, helpers)
    /stores/                   # Zustand stores
    /types/                    # Shared TypeScript types
    /styles/                   # Tailwind config
    ```
  </frontend_stack_defaults>
  <ui_ux_best_practices>
    - Visual Hierarchy: Limit typography to 4-5 font sizes and weights for consistency.
    - Color Usage: Use 1 neutral base (e.g., `zinc`) and up to 2 accent colors.
    - Spacing and Layout: Always use multiples of 4 for padding and margins to maintain visual structure.
    - State Handling: Use skeleton placeholders or `animate-pulse` to indicate data loading states.
    - Accessibility: Use semantic HTML and ARIA roles where appropriate. Favor pre-defined roles over custom attributes.
  </ui_ux_best_practices>
  <code_editing_rules>
```

## Collaborative coding in production: Cursor's GPT-5 prompt tuning

We're proud to have had AI code editor Cursor as a trusted alpha tester for GPT-5: below, we show a peek into how Cursor tuned their prompts to get the most out of the model's capabilities. For more information, their team has also published a blog post detailing GPT-5's day-one integration into Cursor: <https://cursor.com/blog/gpt-5>

### System prompt and parameter tuning

Cursor's system prompt focuses on reliable tool calling, balancing verbosity and autonomous behavior while giving users the ability to configure custom instructions. Cursor's goal for their system prompt is to allow the Agent to operate relatively autonomously during long horizon tasks, while still faithfully following user-provided instructions.

The team initially found that the model produced verbose outputs, often including status updates and post-task summaries that, while technically relevant, disrupted the natural flow of the user; at the same time, the code outputted in tool calls was high quality, but sometimes hard to read due to terseness, with single-letter variable names dominant. In search of a better balance, they set the verbosity API parameter to low to keep text outputs



brief, and then modified the prompt to strongly encourage verbose outputs in coding tools only.

```
Write code for clarity first. Prefer readable, maintainable solutions with clear
```

This dual usage of parameter and prompt resulted in a balanced format combining efficient, concise status updates and final work summary with much more readable code diffs.

Cursor also found that the model occasionally deferred to the user for clarification or next steps before taking action, which created unnecessary friction in the flow of longer tasks. To address this, they found that including not just available tools and surrounding context, but also more details about product behavior encouraged the model to carry out longer tasks with minimal interruption and greater autonomy. Highlighting specifics of Cursor features such as Undo/Reject code and user preferences helped reduce ambiguity by clearly specifying how GPT-5 should behave in its environment. For longer horizon tasks, they found this prompt improved performance:

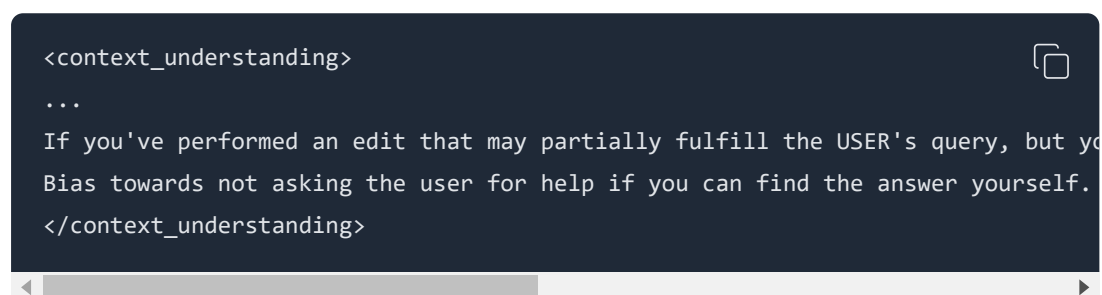
```
Be aware that the code edits you make will be displayed to the user as proposed
```

Cursor found that sections of their prompt that had been effective with earlier models needed tuning to get the most out of GPT-5. Here is one example below:

```
<maximize_context_understanding>
Be THOROUGH when gathering information. Make sure you have the FULL picture before
...
</maximize_context_understanding>
```

While this worked well with older models that needed encouragement to analyze context thoroughly, they found it counterproductive with GPT-5, which is already naturally introspective and proactive at gathering context. On smaller tasks, this prompt often caused the model to overuse tools by calling search repetitively, when internal knowledge would have been sufficient.

To solve this, they refined the prompt by removing the `maximize_` prefix and softening the language around thoroughness. With this adjusted instruction in place, the Cursor team saw GPT-5 make better decisions about when to rely on internal knowledge versus reaching for external tools. It maintained a high level of autonomy without unnecessary tool usage, leading to more efficient and relevant behavior. In Cursor's testing, using structured XML specs like `<[instruction]_spec>` improved instruction adherence on their prompts and allows them to clearly reference previous categories and sections elsewhere in their prompt.



```
<context_understanding>
...
If you've performed an edit that may partially fulfill the USER's query, but you
have found the answer yourself, bias towards not asking the user for help if you
can find the answer yourself.
</context_understanding>
```

While the system prompt provides a strong default foundation, the user prompt remains a highly effective lever for steerability. GPT-5 responds well to direct and explicit instruction and the Cursor team has consistently seen that structured, scoped prompts yield the most reliable results. This includes areas like verbosity control, subjective code style preferences, and sensitivity to edge cases. Cursor found allowing users to configure their own custom Cursor rules to be particularly impactful with GPT-5's improved steerability, giving their users a more customized experience.

## Optimizing intelligence and instruction-following

### Steering

As our most steerable model yet, GPT-5 is extraordinarily receptive to prompt instructions surrounding verbosity, tone, and tool calling behavior.

### Verbosity

In addition to being able to control the `reasoning_effort` as in previous reasoning models, in GPT-5 we introduce a new API parameter called `verbosity`, which influences the length of the model's final answer, as

opposed to the length of its thinking. Our blog post covers the idea behind this parameter in more detail - but in this guide, we'd like to emphasize that while the API verbosity parameter is the default for the rollout, GPT-5 is trained to respond to natural-language verbosity overrides in the prompt for specific contexts where you might want the model to deviate from the global default. Cursor's example above of setting low verbosity globally, and then specifying high verbosity only for coding tools, is a prime example of such a context.

## Instruction following

Like GPT-4.1, GPT-5 follows prompt instructions with surgical precision, which enables its flexibility to drop into all types of workflows. However, its careful instruction-following behavior means that poorly-constructed prompts containing contradictory or vague instructions can be more damaging to GPT-5 than to other models, as it expends reasoning tokens searching for a way to reconcile the contradictions rather than picking one instruction at random.

Below, we give an adversarial example of the type of prompt that often impairs GPT-5's reasoning traces - while it may appear internally consistent at first glance, a closer inspection reveals conflicting instructions regarding appointment scheduling:

- `Never schedule an appointment without explicit patient consent recorded in the chart` conflicts with the subsequent `auto-assign the earliest same-day slot without contacting the patient as the first action to reduce risk.`
- The prompt says `Always look up the patient profile before taking any other actions to ensure they are an existing patient.` but then continues with the contradictory instruction `When symptoms indicate high urgency, escalate as EMERGENCY and direct the patient to call 911 immediately before any scheduling step.`

```
You are CareFlow Assistant, a virtual admin for a healthcare startup that [scheduling]
- Core entities include Patient, Provider, Appointment, and PriorityLevel (Red, Orange, Yellow, Green)
+Core entities include Patient, Provider, Appointment, and PriorityLevel (Red, Orange, Yellow, Green)
*Do not do lookup in the emergency case, proceed immediately to providing 911 guidance
- Use the following capabilities: schedule-appointment, modify-appointment, waitlist-management
- For high-acuity Red and Orange cases, auto-assign the earliest same-day slot
```



- For high-acuity Red and Orange cases, auto-assign the earliest same-day slot

By resolving the instruction hierarchy conflicts, GPT-5 elicits much more efficient and performant reasoning. We fixed the contradictions by:

- Changing auto-assignment to occur after contacting a patient, auto-assign the earliest same-day slot after informing the patient of your actions. to be consistent with only scheduling with consent.
- Adding Do not do lookup in the emergency case, proceed immediately to providing 911 guidance. to let the model know it is ok to not look up in case of emergency.

We understand that the process of building prompts is an iterative one, and many prompts are living documents constantly being updated by different stakeholders - but this is all the more reason to thoroughly review them for poorly-worded instructions. Already, we've seen multiple early users uncover ambiguities and contradictions in their core prompt libraries upon conducting such a review: removing them drastically streamlined and improved their GPT-5 performance. We recommend testing your prompts in our [prompt optimizer tool](#) to help identify these types of issues.

## Minimal reasoning

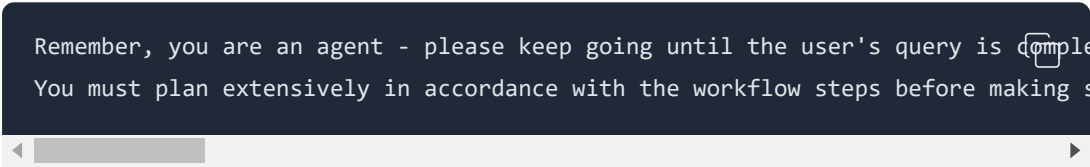
In GPT-5, we introduce minimal reasoning effort for the first time: our fastest option that still reaps the benefits of the reasoning model paradigm. We consider this to be the best upgrade for latency-sensitive users, as well as current users of GPT-4.1.

Perhaps unsurprisingly, we recommend prompting patterns that are similar to [GPT-4.1 for best results](#). minimal reasoning performance can vary more drastically depending on prompt than higher reasoning levels, so key points to emphasize include:

1. Prompting the model to give a brief explanation summarizing its thought process at the start of the final answer, for example via a bullet point list, improves performance on tasks requiring higher intelligence.
2. Requesting thorough and descriptive tool-calling preambles that

continually update the user on task progress improves performance in agentic workflows.

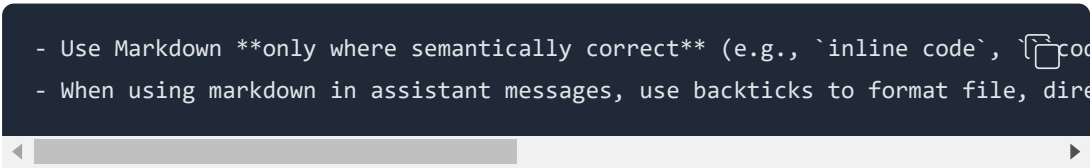
3. Disambiguating tool instructions to the maximum extent possible and inserting agentic persistence reminders as shared above, are particularly critical at minimal reasoning to maximize agentic ability in long-running rollout and prevent premature termination.
4. Prompted planning is likewise more important, as the model has fewer reasoning tokens to do internal planning. Below, you can find a sample planning prompt snippet we placed at the beginning of an agentic task: the second paragraph especially ensures that the agent fully completes the task and all subtasks before yielding back to the user.



```
Remember, you are an agent - please keep going until the user's query is complete. You must plan extensively in accordance with the workflow steps before making s
```

## Markdown formatting

By default, GPT-5 in the API does not format its final answers in Markdown, in order to preserve maximum compatibility with developers whose applications may not support Markdown rendering. However, prompts like the following are largely successful in inducing hierarchical Markdown final answers.



- Use Markdown **only** where semantically correct\*\* (e.g., ``inline code``, ``code``)
- When using markdown in assistant messages, use backticks to format file, directory

Occasionally, adherence to Markdown instructions specified in the system prompt can degrade over the course of a long conversation. In the event that you experience this, we've seen consistent adherence from appending a Markdown instruction every 3-5 user messages.

## Metaprompting

Finally, to close with a meta-point, early testers have found great success using GPT-5 as a meta-prompter for itself. Already, several users have deployed prompt revisions to production that were generated simply by

asking GPT-5 what elements could be added to an unsuccessful prompt to elicit a desired behavior, or removed to prevent an undesired one.

Here is an example metaprompt template we liked:

```
When asked to optimize prompts, give answers from your own perspective - explain  
Here's a prompt: [PROMPT]  
The desired behavior from this prompt is for the agent to [DO DESIRED BEHAVIOR]
```

## Appendix

### SWE-Bench verified developer instructions

```
In this environment, you can run `bash -lc <apply_patch_command>` to execute a  
apply_patch << 'PATCH'  
*** Begin Patch  
[YOUR_PATCH]  
*** End Patch  
PATCH  
Where [YOUR_PATCH] is the actual content of your patch.  
Always verify your changes extremely thoroughly. You can make as many tool calls  
IMPORTANT: not all tests are visible to you in the repository, so even on problems
```

### Agentic coding tool definitions

```
## Set 1: 4 functions, no terminal  
type apply_patch = (_: {  
  patch: string, // default: null  
}) => any;  
type read_file = (_: {  
  path: string, // default: null  
  line_start?: number, // default: 1  
  line_end?: number, // default: 20  
}) => any;  
type list_files = (_: {  
  path?: string, // default: ""  
  depth?: number, // default: 1  
}) => any;
```

```

type find_matches = (_: {
  query: string, // default: null
  path?: string, // default: ""
  max_results?: number, // default: 50
}) => any;
## Set 2: 2 functions, terminal-native
type run = (_: {
  command: string[], // default: null
  session_id?: string | null, // default: null
  working_dir?: string | null, // default: null
  ms_timeout?: number | null, // default: null
  environment?: object | null, // default: null
  run_as_user?: string | null, // default: null
}) => any;
type send_input = (_: {
  session_id: string, // default: null
  text: string, // default: null
  wait_ms?: number, // default: 100
}) => any;

```

As shared in the GPT-4.1 prompting guide, [here](#) is our most updated `apply_patch` implementation: we highly recommend using `apply_patch` for file edits to match the training distribution. The newest implementation should match the GPT-4.1 implementation in the overwhelming majority of cases.

## Taubench-Retail minimal reasoning instructions

```

As a retail agent, you can help users cancel or modify pending orders, return
Remember, you are an agent - please keep going until the user's query is complete
If you are not sure about information pertaining to the user's request, use your tools
You MUST plan extensively before each function call, and reflect extensively on the plan
# Workflow steps
- At the beginning of the conversation, you have to authenticate the user identity
- Once the user has been authenticated, you can provide the user with information
- You can only help one user per conversation (but you can handle multiple requests)
- Before taking consequential actions that update the database (cancel, modify, etc)
- You should not make up any information or knowledge or procedures not provided
- You should at most make one tool call at a time, and if you take a tool call,
- You should transfer the user to a human agent if and only if the request cannot be handled
## Domain basics
- All times in the database are EST and 24 hour based. For example "02:30:00" means 2:30 AM EST
- Each user has a profile of its email, default address, user id, and payment method
- Our retail store has 50 types of products. For each type of product, there are multiple items
- Each product has an unique product id, and each item has an unique item id. The

```

- Each order can be in status 'pending', 'processed', 'delivered', or 'cancelled'
- Exchange or modify order tools can only be called once. Be sure that all items are available

## Cancel pending order

- An order can only be cancelled if its status is 'pending', and you should check that all items are available
- The user needs to confirm the order id and the reason (either 'no longer needed' or 'other')
- After user confirmation, the order status will be changed to 'cancelled', and the order will be refunded

## Modify pending order

- An order can only be modified if its status is 'pending', and you should check that all items are available
- For a pending order, you can take actions to modify its shipping address, payment method, or items

## Modify payment

- The user can only choose a single payment method different from the original one
- If the user wants to modify the payment method to gift card, it must have enough balance
- After user confirmation, the order status will be kept 'pending'. The original payment method will be refunded

## Modify items

- This action can only be called once, and will change the order status to 'pending'
- For a pending order, each item can be modified to an available new item of the same category
- The user must provide a payment method to pay or receive refund of the price difference

## Return delivered order

- An order can only be returned if its status is 'delivered', and you should check that all items are available
- The user needs to confirm the order id, the list of items to be returned, and the reason (either 'no longer needed' or 'other')
- The refund must either go to the original payment method, or an existing gift card
- After user confirmation, the order status will be changed to 'return requested'

## Exchange delivered order

- An order can only be exchanged if its status is 'delivered', and you should check that all items are available
- For a delivered order, each item can be exchanged to an available new item of the same category
- The user must provide a payment method to pay or receive refund of the price difference
- After user confirmation, the order status will be changed to 'exchange requested'

## Terminal-Bench prompt

Please resolve the user's task by editing and testing the code files in your current environment. You are a deployed coding agent.

Your session is backed by a container specifically designed for you to easily run code and test your solutions.

You MUST adhere to the following criteria when executing the task:

<instructions>

- Working on the repo(s) in the current environment is allowed, even if they are not the user's repo(s).
- Analyzing code for vulnerabilities is allowed.
- Showing user code and tool call details is allowed.
- User instructions may overwrite the `_CODING GUIDELINES_` section in this development environment.
- Do not use `\ls -R\`, `\find\`, or `\grep\` - these are slow in large repos.
- Use `\apply_patch\` to edit files: `{"cmd":["apply_patch","*** Begin Patch\n"]}`
- If completing the user's task requires writing or modifying files:
  - Your code and final answer should follow these `_CODING GUIDELINES_`:
    - Fix the problem at the root cause rather than applying surface-level patches
    - Avoid unneeded complexity in your solution.



- Ignore unrelated bugs or broken tests; it is not your responsibility to
- Update documentation as necessary.
- Keep changes consistent with the style of the existing codebase. Changes s
  - Use `\`git log\`` and `\`git blame\`` to search the history of the codebase
- NEVER add copyright or license headers unless specifically requested.
- You do not need to `\`git commit\`` your changes; this will be done automati
- If there is a `.pre-commit-config.yaml`, use `\`pre-commit run --files ...\``
  - If pre-commit doesn't work after a few retries, politely inform the user
- Once you finish coding, you must
  - Check `\`git status\`` to sanity check your changes; revert any scratch fi
  - Remove all inline comments you added much as possible, even if they look
  - Check if you accidentally add copyright or license headers. If so, remov
  - Try to run pre-commit if it is available.
  - For smaller tasks, describe in brief bullet points
  - For more complex tasks, include brief high-level description, use bullet
- If completing the user's task DOES NOT require writing or modifying files (e.
- Respond in a friendly tune as a remote teammate, who is knowledgeable, capab
- When your task involves writing or modifying files:
  - Do NOT tell the user to "save the file" or "copy the code into a file" if yo
  - Do NOT show the full contents of large files you have already written, unles

</instructions>

<apply\_patch>

To edit files, ALWAYS use the `\`shell\`` tool with `\`apply_patch\`` CLI. `\`apply`  
`\`\`\`bash`  
`{"cmd": ["apply_patch", "<<'EOF'\n*** Begin Patch\n[YOUR_PATCH]\n*** End Pat`  
`\`\`\``

Where [YOUR\_PATCH] is the actual content of your patch, specified in the follow  
 \*\*\* [ACTION] File: [path/to/file] -> ACTION can be one of Add, Update, or Delet

For each snippet of code that needs to be changed, repeat the following:

[context\_before] -> See below for further instructions on context.

- [old\_code] -> Precede the old code with a minus sign.

+ [new\_code] -> Precede the new, replacement code with a plus sign.

[context\_after] -> See below for further instructions on context.

For instructions on [context\_before] and [context\_after]:

- By default, show 3 lines of code immediately above and 3 lines immediately be

- If 3 lines of context is insufficient to uniquely identify the snippet of cod

```
@@ class BaseClass
```

```
[3 lines of pre-context]
```

```
- [old_code]
```

```
+ [new_code]
```

```
[3 lines of post-context]
```

- If a code block is repeated so many times in a class or function such that ev

```
@@ class BaseClass
```

```
@@ def method():
```

```
[3 lines of pre-context]
```

```
- [old_code]
```

```
+ [new_code]
```

```
[3 lines of post-context]
```

Note, then, that we do not use line numbers in this diff format, as the context is preserved.

```
\\\`bash
{"cmd": ["apply_patch", "<<'EOF'\n*** Begin Patch\n*** Update File: pygorithm\n\n"]}
```

File references can only be relative, NEVER ABSOLUTE. After the apply\_patch command, use the following tags to structure your response:

```
</apply_patch>
<persistence>
```

You are an agent - please keep going until the user's query is completely resolved.

- Never stop at uncertainty – research or deduce the most reasonable approach and document it.
- Do not ask the human to confirm assumptions – document them, act on them, and document the results.

```
</persistence>
<exploration>
```

If you are not sure about file content or codebase structure pertaining to the task, use the following tags to structure your response:

Before coding, always:

- Decompose the request into explicit requirements, unclear areas, and hidden assumptions.
- Map the scope: identify the codebase regions, files, functions, or libraries involved.
- Check dependencies: identify relevant frameworks, APIs, config files, data formats, etc.
- Resolve ambiguity proactively: choose the most probable interpretation based on context.
- Define the output contract: exact deliverables such as files changed, expected behavior, etc.
- Formulate an execution plan: research steps, implementation sequence, and testing strategy.

```
</exploration>
<verification>
```

Routinely verify your code works as you work through the task, especially any changes to existing code.

Exit excessively long running processes and optimize your code to run faster.

```
</verification>
<efficiency>
```

Efficiency is key. you have a time limit. Be meticulous in your planning, tool usage, and code quality.

```
</efficiency>
<final_instructions>
```

Never use editor tools to edit files. Always use the `apply\_patch` tool.

```
</final_instructions>
```