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GAIA2: BENCHMARKING LLM AGENTS ON *DYNAMIC* AND *ASYNCHRONOUS* ENVIRONMENTS

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The "Sim-to-Real" Gap in Agent Evaluation

Current limitations

- Most benchmarks (e.g. GAIA, SWE-Bench, TerminalBench, Appworld, ...) are **synchronous and agent-driven**.
- Environments are **strictly reactive** and lack independent temporal dynamics.
- Missing key elements to test real-world deployments failure modes (temporal reasoning, situational awareness, ...).

Dynamic Environments

- **Asynchronous & Event-driven**.
- Agents face **temporal constraints, background noise, ambiguity**, ...

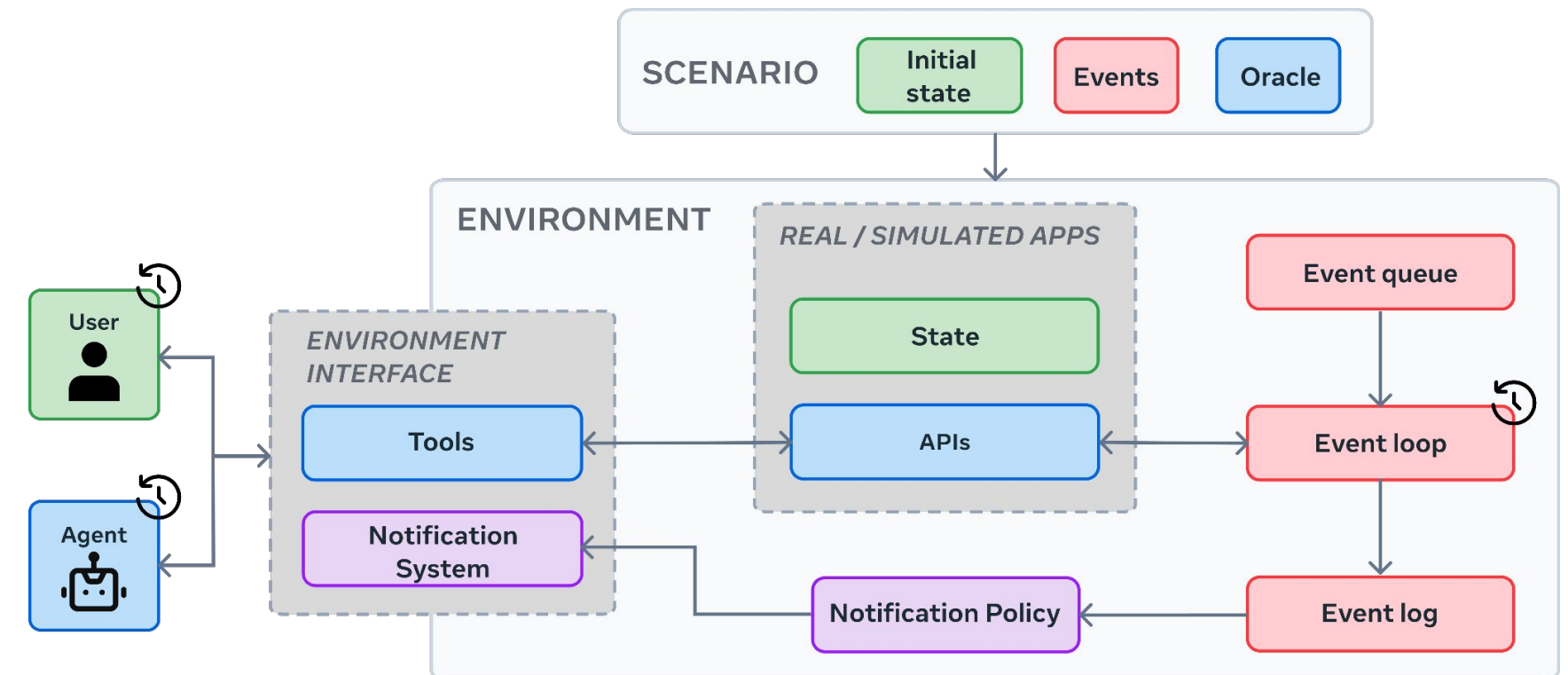
The Enabler: Agents Research Environments (ARE)

- A general-purpose **simulation platform** where time flows continuously and independently of the agent .
- ARE provides the foundation for **GAIA2: our new benchmark for General Agents in dynamic environments**.


Agents Research Environments

Core Abstractions:

1. **Apps:** Stateful APIs with read/write tools
2. **Environment:** Time-indexed collection of apps.
3. **Events:** The user receives an email from a friend as an external event, which notifies the Agent. **Everything is an event:** logged into a single EventLog. Events are either “read” or “write” (modify the user space).
4. **Notifications:** Observability layer that pushes relevant events to the agent.
5. **Scenarios:** Dynamic tasks with verifiable goals.



GAIA2: 1,120 scenarios for Personal General Agents



Mobile Environment (ARE Instantiation)

12 apps • 101 tools • 10 universes

Contacts, Email, Calendar, Messages, Shopping, ...

- 400K-800K tokens per universe
- Persona-driven, cross-app consistent data

A Personal, API-Driven World

- User-Centric Universes:** Each instance is a complete smartphone environment centered around a specific user.
- Rich Personal Content:** Populated with cross-app consistent data (e.g., contacts match messaging history and calendar events).
- Simple Agent Scaffold:** Agents run with a simple ReAct loop where they can perform a single tool call per turn

Capability	Example Task	Explanation
Execution	<i>Update all my contacts aged 24 or younger to be one year older than they are currently</i>	Evaluates the ability to chain long seq. of write actions in the right order
Search	<i>Which city do most of my friends live in? In case of a tie, return the first city alphabetically</i>	Evaluates the ability to chain long seq. of read actions in the right order
Ambiguity	<i>Schedule a 1h Yoga event each day at 6:00 PM from October 16, 2024 to October 21, 2024</i>	Tests whether agents ask for clarification on impossible, contradictory, or ambiguous tasks
Adaptability	<i>I have to meet my friend Kaida to view a property [...] If she replies to suggest another property or time, update the calendar event</i>	Requires agents to adapt dynamically to environmental changes
Time	<i>Send messages to each of the colleagues I am supposed to meet today, asking who is supposed to order the cab. If after 3 minutes there is no response, order a cab from [...]</i>	Evaluates whether agents can complete tasks in due time & maintain temporal awareness
Agent2Agent	<i>*Same Search task as above but the Contacts and Chats apps are replaced by app sub-agents*</i>	Tests whether agents can collaborate with other agents to use tools & complete tasks
Noise	<i>*Same Adaptability task as above but with random tool execution errors and random environment events occurring during execution*</i>	Evaluates whether agents are robust to environment noise & distractors

Scenarios as DAGs of Events

Scenario annotations can be time-indexed to schedule events at specific timestamps.

Time Scenario Task:

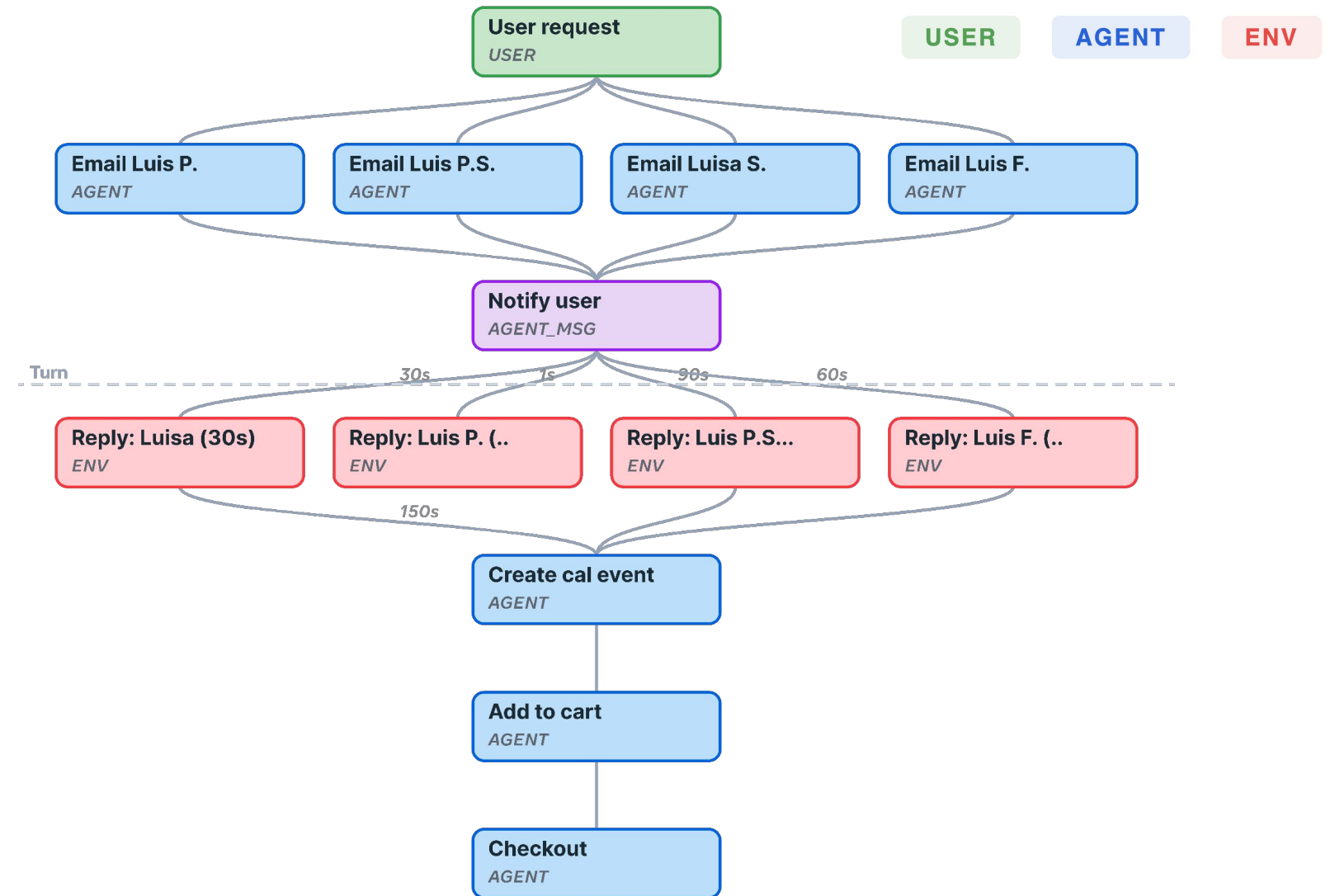
“Email my Friends to **ask for their availability**, and **wait for their reply** to choose a Party time slot that fits in everyone’s calendar”

Environment:

Friends reply to the email with different delays

Challenge:

Capture email **notifications**, **wait** for all replies and then **book accordingly**.



Time Scenario DAG are time-indexed

Anatomy of the Grader

RQ1: What do we grade?

- Agents can do unlimited read actions (exploration).
- Only write actions (environment-altering) are graded using the `EventLog`.

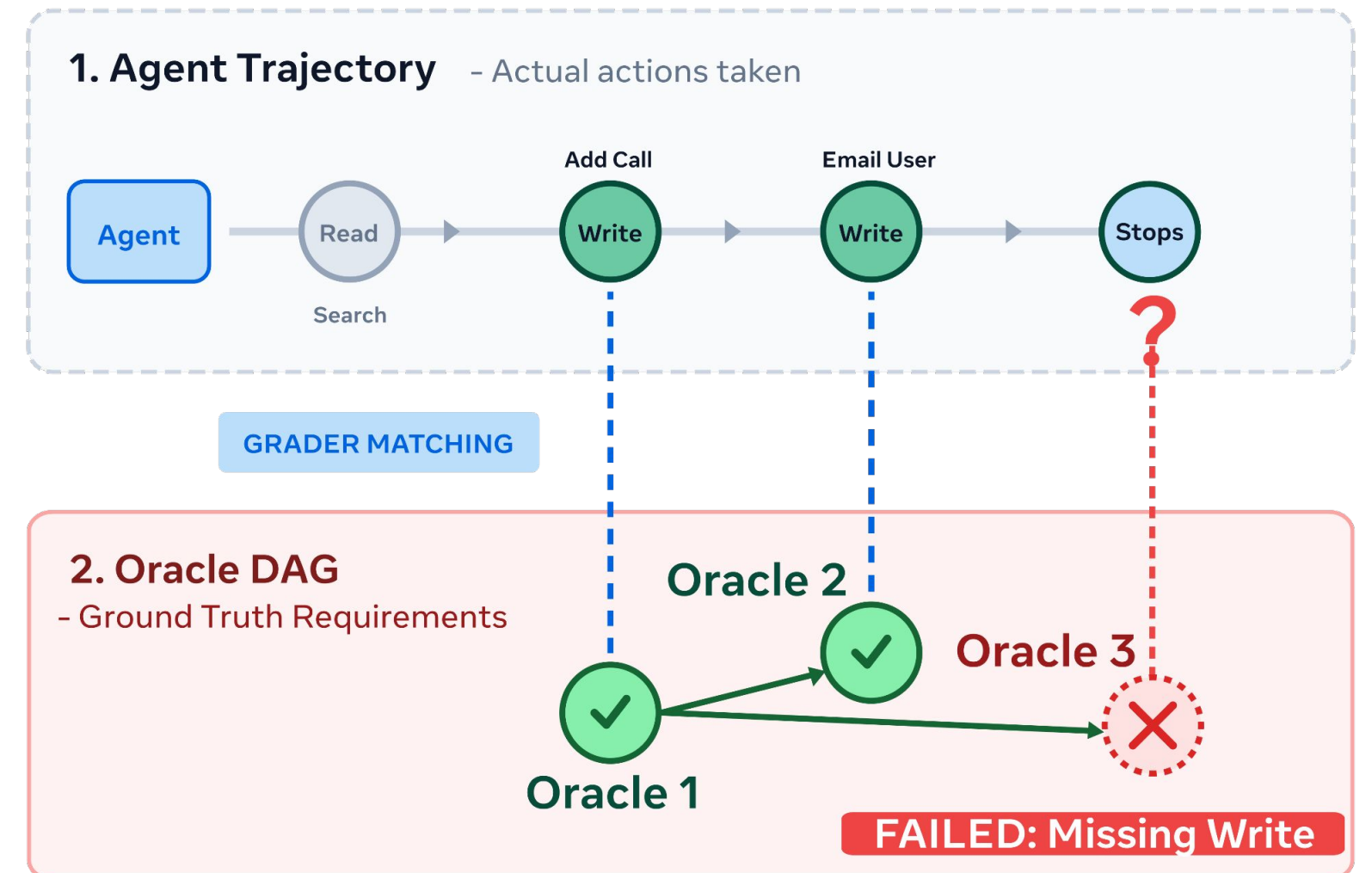
RQ2: How to grade long trajectories?

- Remap the linear `EventLog` to the DAG Oracle annotations.
- 4 pillars of the grader:
- **Consistency**: Exact match / LLM rubrics
 - **Causality**: Enforce DAG dependency
 - **Timing**: Tolerance windows for timely actions
 - **Completeness**: Matching all Oracle actions

RQ3: Can we beat a baseline LLM-as-Judge?

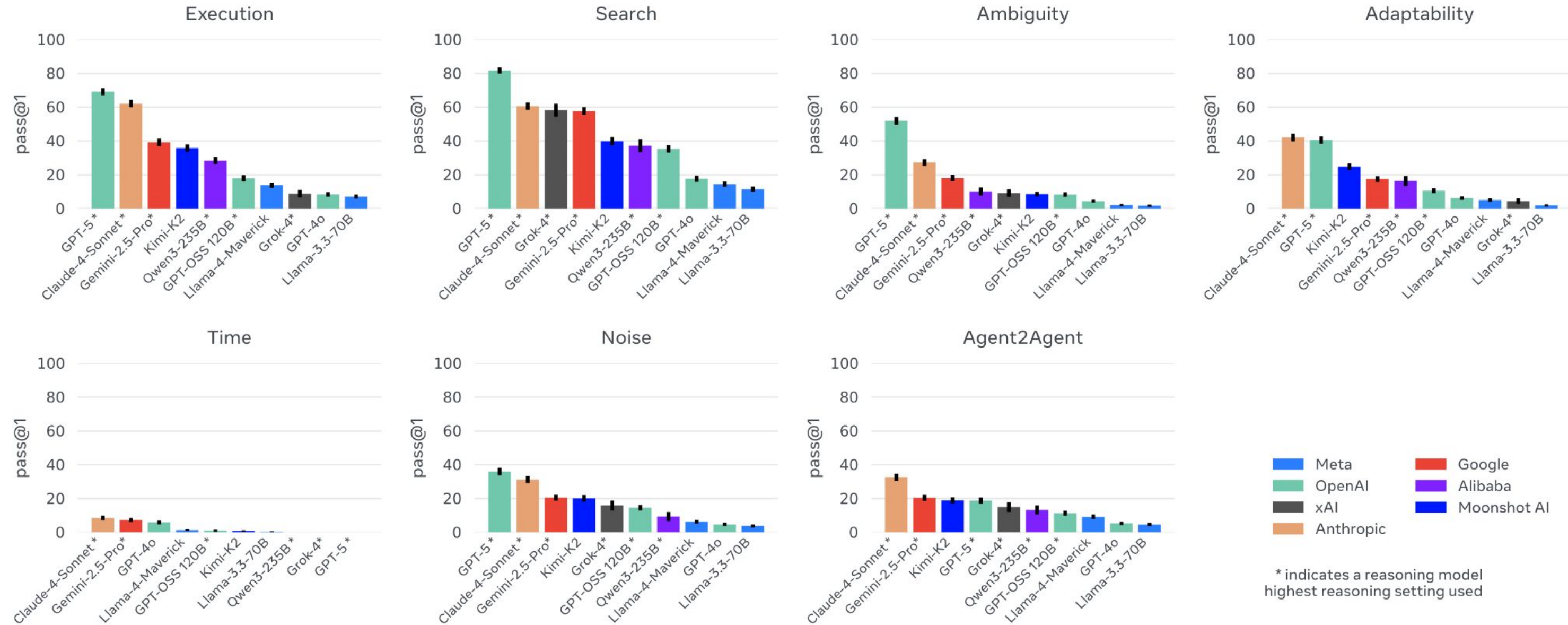
- We compared our DAG-based grader with a pure LLM-as-Judge on a held-out trace set.
- Result: The DAG structure adds significant robustness, preventing context-overload and hallucinations.

GRADER MODEL	AGREEMENT	PRECISION	RECALL
LLM Judge	0.72	0.53	0.83
✓ ARE Grader (Ours)	0.98	0.99	0.95



Main Results

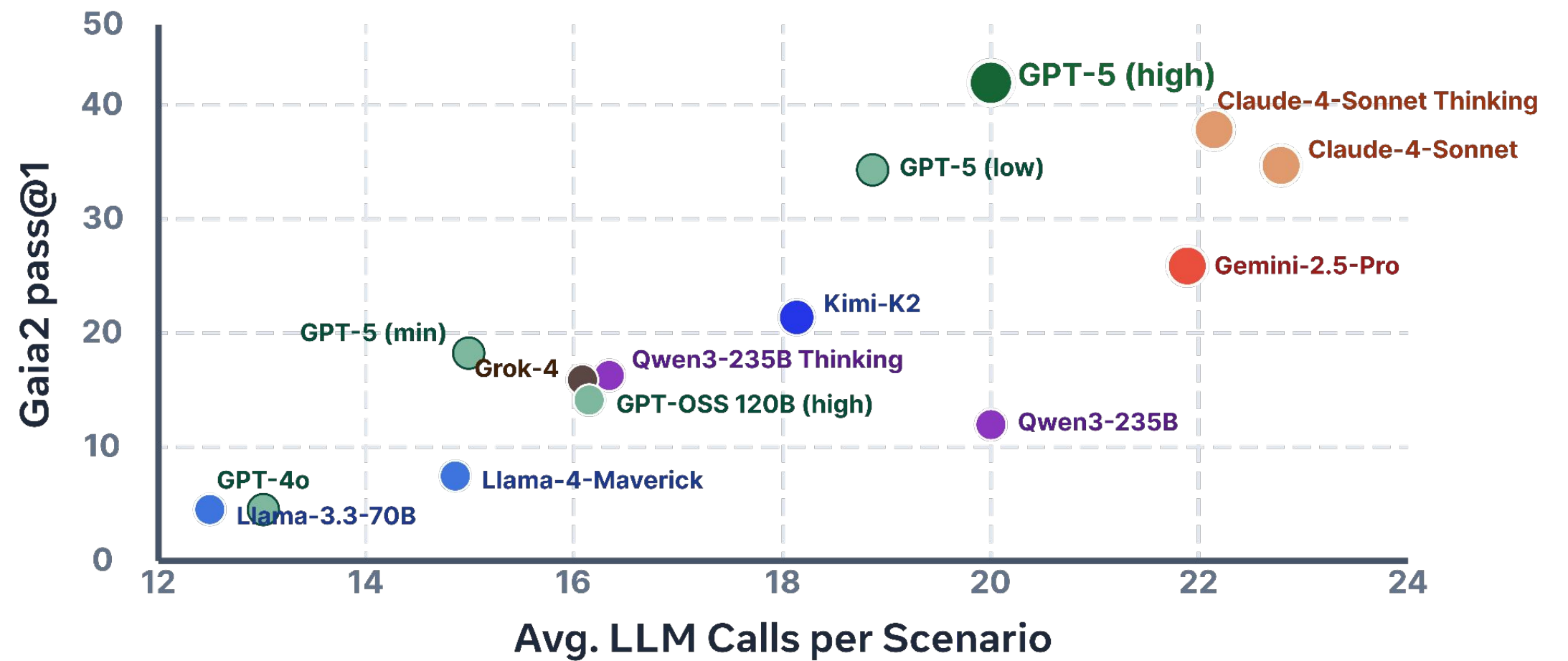
1. Frontier closed-source models dominate across the board.
2. Execution and Search are almost already saturated.
3. Room for improvement on all other capabilities, required in real Agents deployment.



* indicates a reasoning model highest reasoning setting used

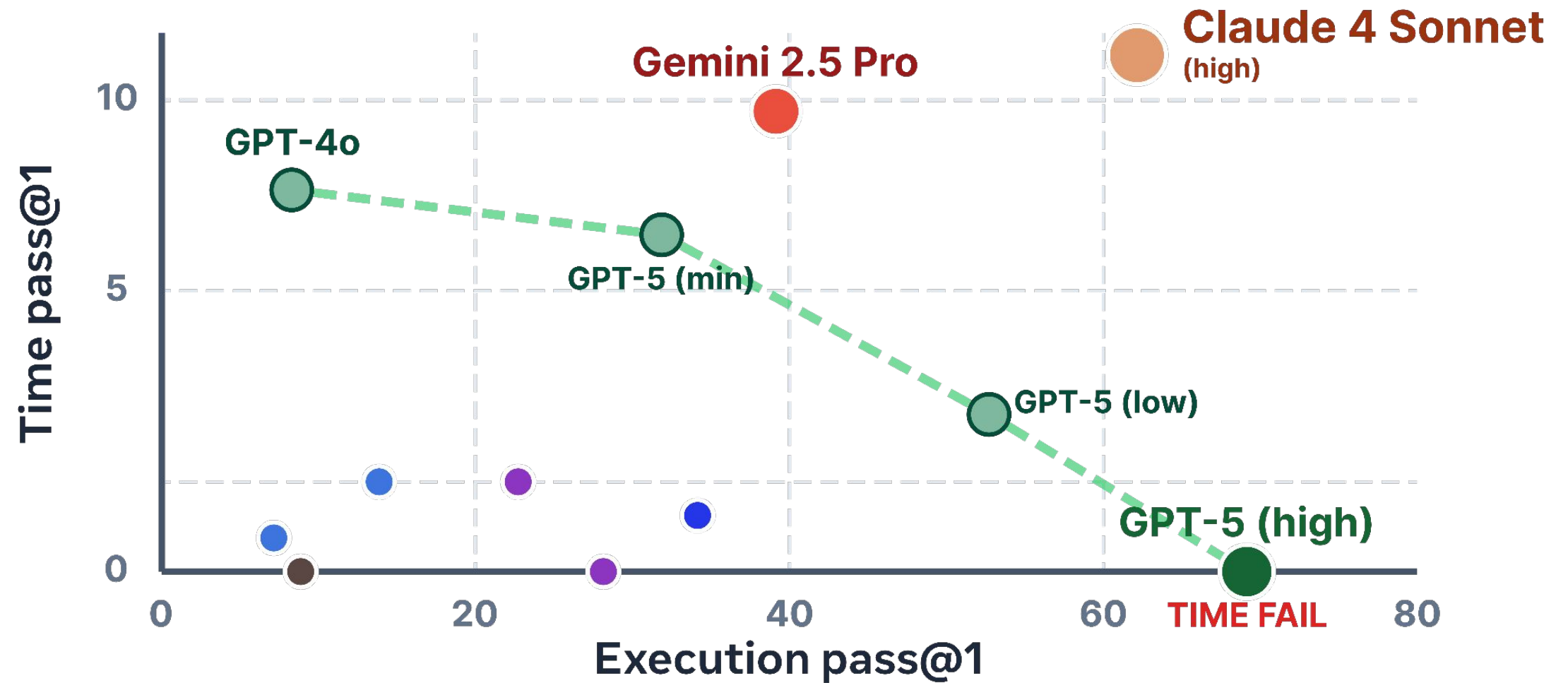
Deeper Search Drives Performance

- **Scale with Effort:** Overall performance scales linearly with the number of average tool calls per scenario.
- **The "Give Up" Factor:** Weaker models tend to stop after just a few failed calls.
- **Read Before You Write:** Strong models (e.g., GPT-5, Claude-4-Sonnet) execute significantly more read calls before making their first write action.
- **Takeaway:** Slow and steady wins the trace.



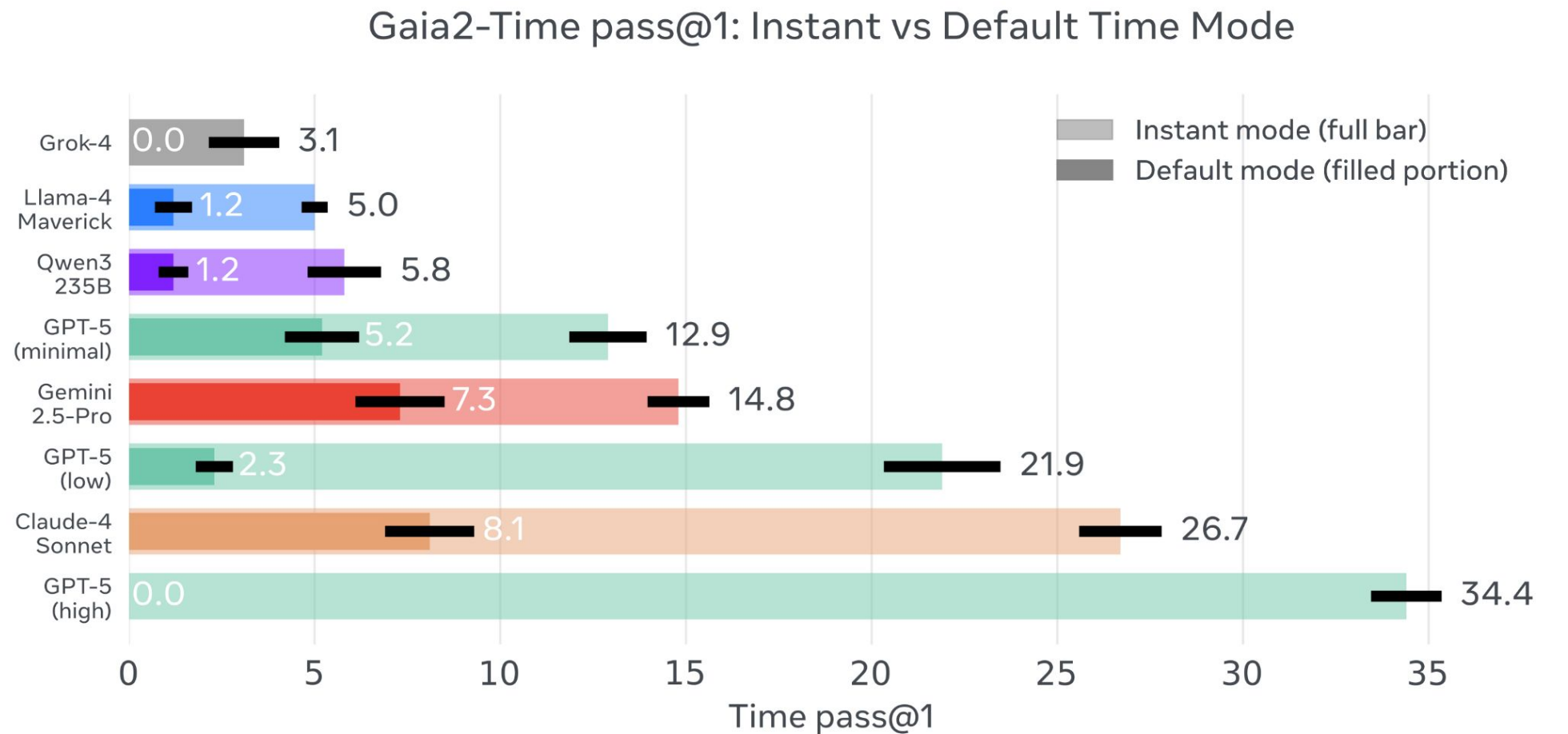
The "Inverse Scaling" Trap on Time Tasks

- **The Deliberation Cost:** Top models use extensive compute during inference, resulting in slow execution.
- **The Penalty:** Heavy reasoning models (like the GPT-5 family) frequently miss deadlines.
- **Takeaway:** An observed inverse scaling law; high execution pass@1 correlates with high failure rates on strict temporal tasks.



The Need for Adaptive Compute

- **The "Instant Time" Ablation:** By freezing time during agent inference in ARE, we isolate the impact of reasoning speed.
- **Capability vs. Speed:** Models have the reasoning capability to solve time-aware tasks (cf.'Instant' mode), but fail in real-time due to slow inference.
- **Takeaway:** Future agents require adaptive compute; fast generation for urgent/routine tasks, and deep reasoning only when the environment permits.



Conclusion

Key Takeaways:

- **The "Sim2Real" Gap is Real:** No single model dominates. SOTA (GPT-5) caps at 42% pass@1.
- **Crucial Trade-offs:** Thinking models perform well when there is no time deadline.
- **Action-Level Verification:** End-state checks are insufficient. The ARE Verifier ensures highly accurate verification and prevents approving models with *destroy* and *repair* behaviors.

What's Next:

- **ARE and GAIA2 provide the open infrastructure** needed to train the next generation of personal agents à la OpenClaw.



Read the full paper



Code & Datasets

References

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