ANTHROP\C

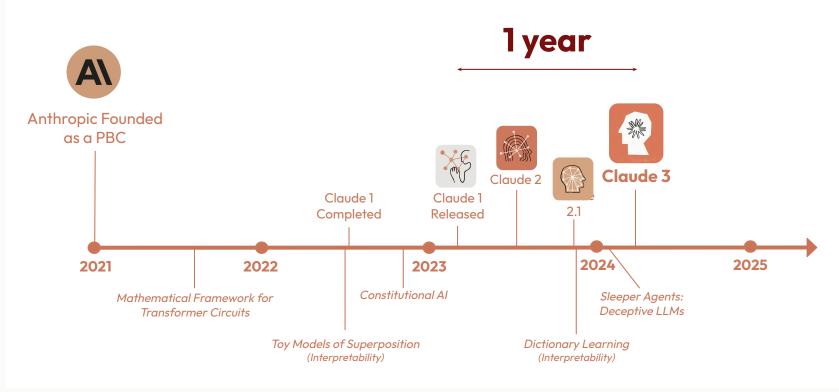
Measuring Agent Capabilities and Anthropic's RSP

Berkeley CS294/194-196 Large Language Model Agents

Add your questions!

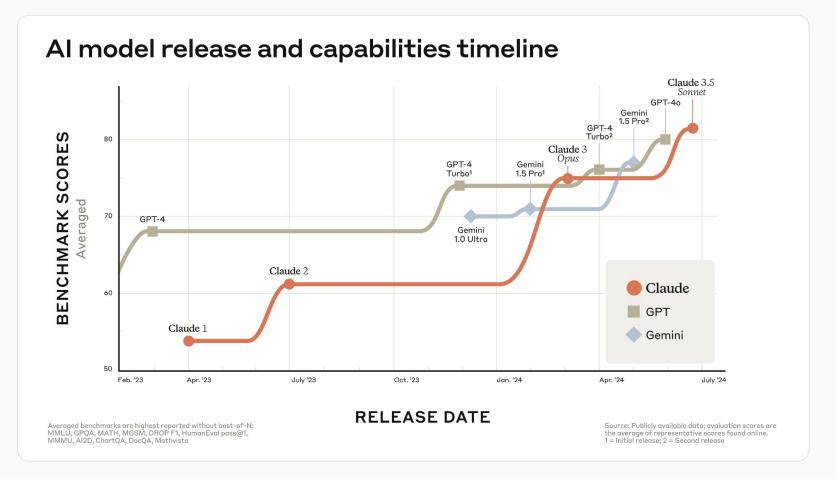


Anthropic's history



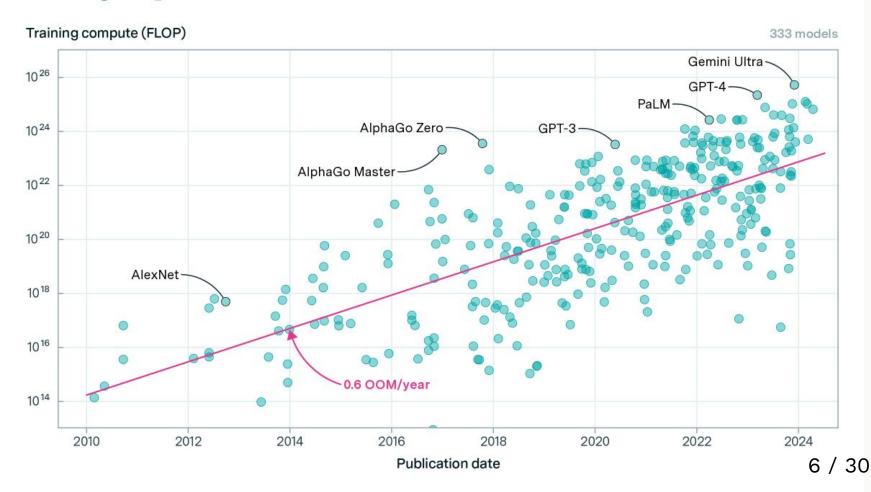
Overview

- Why measurement matters
- Our approach to AI development
- Themes: agents, safety, evaluation

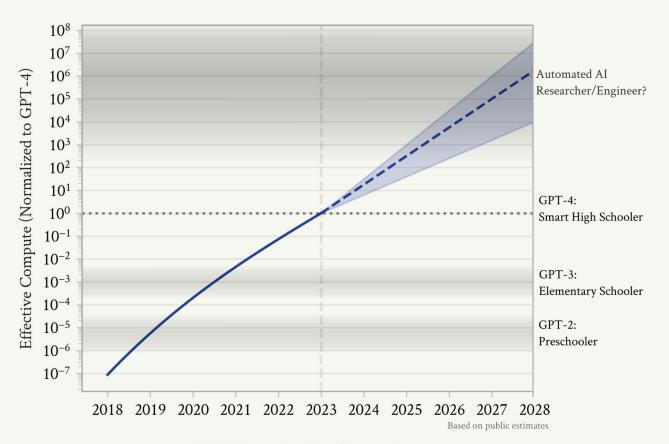


Training compute of notable models



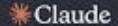


Base Scaleup of Effective Compute

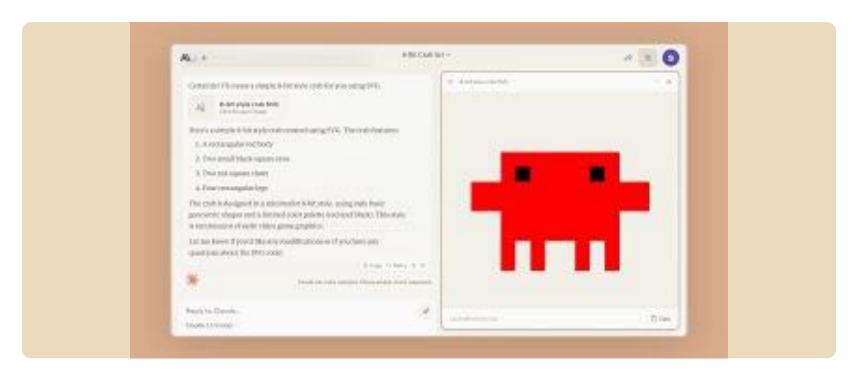


Claude 3.5 Sonnet / Agentic coding

Agentic coding with Claude 3.5 Sonnet



Claude 3.5 Sonnet / Game development



Responsible Scaling Policy

How is Anthropic developing safe models?

Safety frameworks

- Historical context
- Existing parallels
- Anthropic's RSP
- Updates

What is the RSP?

The Responsible Scaling Policy represents Anthropic's public commitment to ensuring that model capability does not outstrip our ability to create effective guardrails for that capability and mitigate harm.

Our Responsible Scaling Policy outlines how we will measure for potential catastrophic risks and then mitigate them

Our goals are to:

- Provide structure to help us make hard decisions about safety
- Hold ourselves publicly accountable to developing models safely
- Learn how to make and **iterate** on safe decisions
- Provide a template for policymakers and others in industry

AI Safety Levels

High Level Overview of AI Safety Levels (ASLs)

ASL-1 (smaller models)

ASL-2 (present large models)

ASL-3 (significantly higher risk)

ASL-4 (speculative)

Increasing Effective Compute









How do we implement our RSP?

When Anthropic approaches a new level of model capability, the RSP mandates that Anthropic prepare necessary safety measures for it.

We are preparing for ASL-3.

The RSP highlights AI Safety Levels:

ASL-1
Smaller Models

ASL-2
Present Large
Models

ASL-3
Significantly
Higher Risk

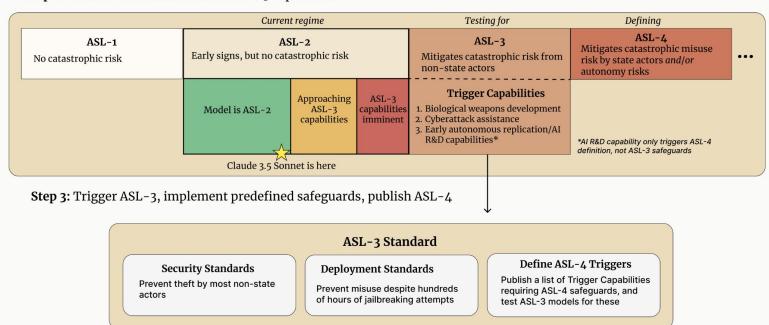
ASL-4
Speculative

ANTHROP\C

How do we implement our RSP?

Step 1: Group capability triggers and corresponding safeguards by AI Safety Level (ASL)

Step 2: Evaluate current models for ASL-3 capabilities



Measuring capabilities

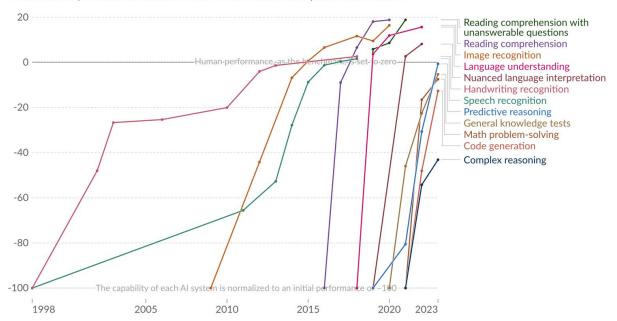
- Challenges
- Anthropic's methods
- Case study: computer use

Benchmarks don't last

Test scores of AI systems on various capabilities relative to human performance



Within each domain, the initial performance of the Al is set to -100. Human performance is used as a baseline, set to zero. When the Al's performance crosses the zero line, it scored more points than humans.

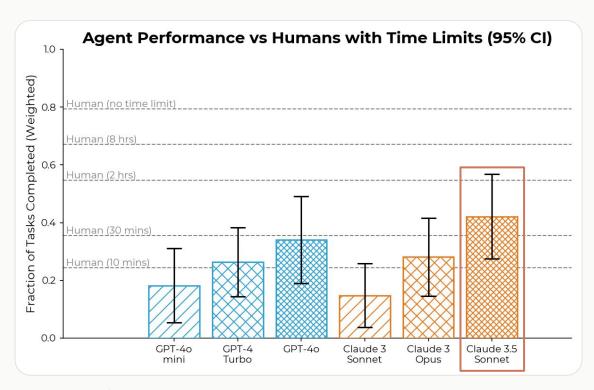


Test scores of AI systems on various capabilities relative to human performance. Human performance is used as a baseline, set to zero.

Kiela et al. (2023)

Measuring task completion time relative to humans

Claude 3.5 Sonnet can handle tasks that would take human developers around 30 minutes, in seconds.



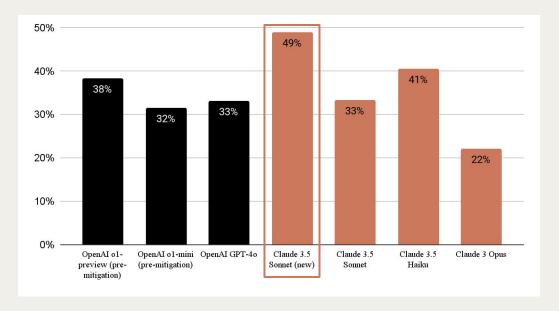
August 2024 Eval from METR (Model Evaluation and Threat Research)

3.5 Sonnet outperforms OpenAl ol

Despite being significantly faster and cheaper than o1*, Claude 3.5 Sonnet beats o1-preview on key agentic benchmarks like SWE-bench Verified – making it the best choice across the industry for production usage

SWE-bench Verified

Assesses a model's ability to complete real-world software engineering tasks and understand, modify, and test code with tools



Sources: OpenAI of model system card, Anthropic model card (October 2024 3.5 model addendum)

^{*}Claude 3.5 Sonnet is \$3/Mtok Input and \$15/Mtok Output, compared to OpenAI o1 at \$15/Mtok Input and \$60/Mtok output (billing is also non-deterministic)

3.5 Sonnet outperforms OpenAl ol

Despite being significantly faster and cheaper than o1*, Claude 3.5 Sonnet beats o1-preview on key agentic benchmarks such the aider code editing benchmark – making it the practical choice

Aider's code editing benchmark

Measures the LLM's coding ability, ability to write new code that integrates into existing code, and apply changes to the source file without human intervention



Sources: Aider LLM Leaderboards

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3.5 Sonnet outperforms OpenAl ol

Despite being significantly faster and cheaper than o1*, Claude 3.5 Sonnet beats o1-preview on key agentic benchmarks such as an 18% lead in the aider refactoring benchmark – making it the practical choice

Aider's refactoring benchmark

A more challenging benchmark which tests the model's ability to output long chunks of code without skipping sections or making mistakes



Sources: Aider LLM Leaderboards

^{*}Claude 3.5 Sonnet is \$3/Mtok Input and \$15/Mtok Output, compared to OpenAI o1 at \$15/Mtok Input and \$60/Mtok output (billing is also non-deterministic)

Case study

Computer use

Computer use

- Demos!
- Technical implementation
- Safety considerations
- Future implications

Claude 3.5 Sonnet (upgraded)

Computer use demo: Coding



Claude 3.5 Sonnet (upgraded)

Computer use demo: Operations



Practical safety measures

- ASL standard implementation
- Security measures
- Deployment safeguards
- Lessons learned in year 1

Future

- Scaling governance
- Capability measurement
- Academic collaboration

Add your questions!



Thank you A