TACI.ai | TASK-AI CAPABILITY INDEX CONFIDENTIAL ADVISOR BRIEF | v0 .1 | JULY 2025

MEASURING LLM JOB-TASK PERFORMANCE: 180-Task Core Slice,

24-Hour Turn-Around

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<u>Abstract</u>

TACI benchmarks frontier language models on a 180-task slice of real ONET job statements that spans text, GUI, and vision. Each run finishes within twenty-four hours of a new checkpoint, with SHA-256-logged prompts, six-axis grading, and 95 % confidence intervals. The framework is adaptable: any occupation from the full ≈14 000-task ONET taxonomy can be analysed by selecting the desired rows. This brief shows one occupation (Paralegals) and invites methodological feedback from researchers, model labs, and business decision-makers.

TABLE OF CONTENTS

- 1. Why This Matters
- What Makes TACI Different ★Your Input Requested
- 3. Spotlight Paralegals & Legal Assistants
- 4. Method in Four Steps
- 5. Comparative Context
- 6. Contact & Next Steps

1. WHY THIS MATTERS

Public model leaderboards still test synthetic trivia and academic exams. Businesses, Al labs, and policymakers need to know how new checkpoints perform on the *actual work tasks* that power the economy.

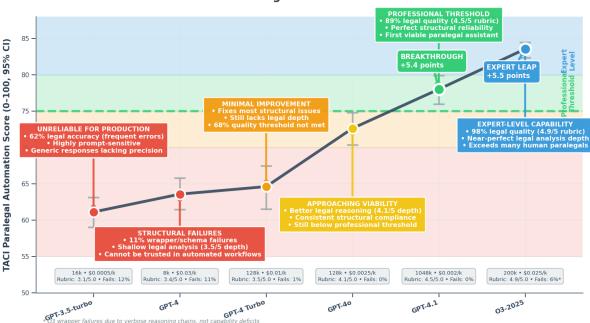
TACI answers that gap with a reproducible, twenty-four-hour pipeline over a 180-task core slice. Because every O*NET record is indexed, the system can be directed at **any occupation** in the U.S. Department of Labor taxonomy, today, by toggling the input manifest.

2. WHAT MAKES TACI DIFFERENT

- Real-task grounding: Importance-weighted O*NET tasks.
- Multi-modal prompts: Text, structured GUI actions, vision findings.
- 24-h refresh cadence: New checkpoint, new numbers next day.
- Transparent hashing: Prompts, outputs, and grades SHA-256-logged.
- Six-axis rubric + hard gates: Quality and structural reliability.
- Statistical discipline: 1 000-bootstrap 95 % confidence intervals.
- ★ Your input requested A 30-minute call to critique design choices would be invaluable. If the approach proves useful, may we list you as an informal advisor (name & head-shot; no ongoing duties)?

3. SPOTLIGHT - PARALEGALS & LEGAL ASSISTANTS (SOC 23-2011)

Line chart of TACI paralegal scores for six GPT checkpoints; annotations mark structural failures, viability, expert level.



Al Legal Assistant Evolution: From Unreliable to Expert-Level
TACI Paralegal Benchmark Results

Each point aggregates all 14 O*NET paralegal tasks with task-level importance weights. Results reflect 72 outputs per model (14 tasks × 2 temperatures × 3 prompt variants). Error bars show 95% confidence intervals computed from 1,000 bootstrap samples.

- Reliability Cliff: The 11% Structural Failure Problem. GPT-4 has an 11% wrapper/schema failure rate, making it fundamentally unreliable for automated legal workflows, while GPT-4.1+ achieve perfect structural compliance. This is not just a performance gap; it's a trust threshold for real deployment.
- Professional Viability Jump: GPT-4.1 as the Inflection Point. TACI pinpoints when AI becomes professionally viable: GPT-4.1 scores 78% with 4.5/5 rubric quality, crossing the "colleague-level" threshold. Below this, AI is only assistance-level; above, it can do independent paralegal work.
- Expert Paradox: O3's Tradeoff: Higher Quality, More Failures. O3 delivers 98% legal quality (4.9/5) and 83.5% score, despite 6% wrapper failures. GPT-4.1 has 0% failures

but lower quality. This shows reasoning depth and formatting compliance are separate axes: O3's verbose reasoning boosts analysis but breaks wrappers.

The identical analysis can be generated for *any* occupation or custom task list within twenty-four hours.

4. METHOD IN FOUR STEPS

Manifest Ingest O*NET rows → tag modality
 → MD5 hash
 Prompt Matrix Three variants × two temperatures; JSON wrapper
 Batch Run ~1 080 prompts per model (Claude Opus 4, GPT-o3, Llama-4, etc.
 Layered Grading Wrapper → JSON schema → Safety → six-axis rubric → 1 000× bootstrap CIs

Context-and-price bonus (v0 .1): Extra weight for models offering \geq 128 k context at \leq \$0.002 / k tokens. (adjustable)

5. COMPARATIVE CONTEXT

Dimension	HELM (Stanford '22)	MMLU (OpenAl '21)	BIG-Bench (Google '22)	TACI (2025)
Real-world	48 evaluation	57	399 synthetic	≈14 000 O*NET
grounding	scenarios	academic	tasks (BIG-	job tasks
	(HELM v2.1)	MCQs	Bench full)	
Modalities	Text	Text	Text	Text + GUI +
				Vision
Refresh	Manual	Frozen	One-off release	24-h automated
cadence	releases	snapshot		core slice
Longitudinal	X	X	X	Built-in panel
deltas				tracking
Main use-	Ethics research	Education	Benchmark	Labour policy
cases		testing	research	Business strategy Model
				eval

HELM v2.1 (public release Oct 2024) contains 48 scenarios; the original 2022 paper listed 42. † BIG-Bench figure uses the full 399task repository; many summaries cite the 204-task "Lite" subset.

6. CONTACT & NEXT STEPS

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To discuss or request additional occupations, please choose any open slot:

https://calendly.com/founder-taci/30min