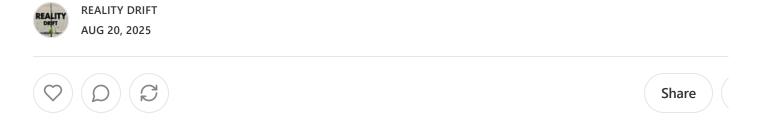
Semantic Drift: The Blindspot AI Researchers Keep Missing

The hidden reason GPT-5 feels blander than GPT-4



This essay is Part 1 in my ongoing series on "Semantic Drift", where I explore how AI resha language, meaning, and culture.

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Cogito, ergo sum

Untiquam te labatio ir empta es consocliax luctar it que ringal injudentiae interreris legt ces bidiciant ariomus obstaxusmod prophecnia et pliasit phrasmir compressino variamce - corcaps

The Distinction

Current benchmarks focus on factual accuracy and hallucinations. But new evidenc suggests another failure mode. Semantic drift: where outputs remain factually correbut lose the original purpose or intent.

Example: Descartes' "Cogito, ergo sum" recast as leadership advice about confidence Factually fine, semantically hollow.

The Metric

We call this *purpose fidelity:* the degree to which AI preserves the meaning, context, and intent of source material. Early experiments show that semantic fidelity degrad far faster than factual accuracy over recursive generations.

Why Drift Happens

Semantic drift isn't random. It emerges from three converging forces:

Training Bias: pretraining on dominant narrative forms (e.g., explanatory or busine oriented text) nudges outputs into those grooves.

Safety Smoothing: fine tuning pushes models toward "safe" generalities, often flattening nuance.

User Convergence: most users lean on default prompts, reinforcing predictable phrasing and compressing variance.

Together, these create a pipeline from originality, to compression, to semantic collapse.

The Two Paths

For most users, this means convergence: voices and ideas flatten into sameness.

But early signs suggest a minority who approach AI as a thinking partner rather the shortcut. They generate expansion instead: new metaphors, new language, new thought patterns. (One hypothesis: cognitive diversity, including neurodivergence, play a role. But this requires testing.)

Why It Matters to Al Companies

Benchmarks miss it: Your evals show models "working" while meaning silently collapses.

Adoption risk: If users sense outputs feel hollow, trust erodes.

Differentiation risk: Companies that solve drift will own the narrative of "authenti AI."

Epistemic liability: Recursive retraining on semantically drifted outputs risks long-term model integrity.

What to Track

A Drift Index: monitoring Purpose Fidelity across domains and over recursive generations.

What to Build

Interfaces that surface intent, not just output.

"Friction by design" to disrupt over-compression.

Adaptive pluralism: multiple stylistic/semantic modes rather than a single flattened voice.

Framing Line for Execs

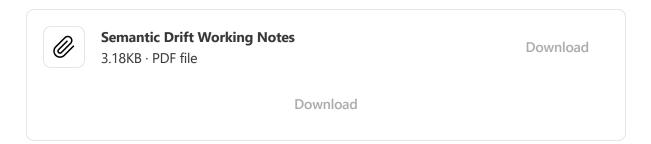
Benchmarks measure models. Drift measures users. If you're not measuring drift, you're flying blind.

Implications

Ignore drift, and you risk flooding the ecosystem with factually correct but semantically hollow text. Solve drift, and you not only protect epistemic stability bu unlock new forms of human-AI co-thinking.

If AI outputs feel hollow, you're not imagining it.

Subscribe for why.



Further Resources:

[The Next Blindspot in AI Evaluation] - Medium

[Semantic Drift PDF Archive] - Offbrandguy

[Semantic Drift Full Documentation] - Zenodo

[Semantic Drift Research Notes] - Figshare



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