

Artificial Intelligence in a Human World

The AI's job is plausibility, not truth.

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The AI said it compiles, the compiler says no.

Hallucination is the feature, not a bug.

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Unless it is factually wrong... then it's a bug.

Plausible (adj.) – *Content that appears convincing, logical, or reasonable to a human observer, and is truthfully accurate, regardless of its objective truth.*

A controversial statement.

Facts make *statements true*

Facts are Truth-bound

A "statement of fact" is a claim that can be proven true or false, *in this context*.

“*Visible Ink*”

Governance: what the system can trust/act on; *deterministic, judge, receipts*

Beliefs make *stories true*

Plausible is Story-bound

“Rewrite this email in a friendlier tone
without changing the meaning.”

“*Invisible Ink*”

Epistemics: what the model can say;
storybound, context-limited

Governance vs. Epistemics

Fact: *Visible Ink*

- True/False
- 0 or 1
- Tools
- A Fact
- Determinism

Plausible: *Invisible Ink*

- Probability
- 0→1
- Rules
- A Story
- Non-determinism

Governance: what the system can trust/act on; *deterministic, judge, receipts*

Epistemics: what the model can say; *storybound, context-limited*

Governance vs. Epistemics

Fact: *Visible Ink*

- “Is Right” vs. “Sounds right”
- “Is Valid” vs. “Looks valid”
- “Correct” vs. “Convincing”

Plausible : *Invisible Ink*

- In Disney, animals can talk
- In DC, Superman can fly
- In TLoR, the one ring is evil

Governance: what the system can trust/act on; *deterministic, judge, receipts*

Epistemics: what the model can say; *storybound, context-limited*

*If a superman movie turned into a batman movie, it wouldn't
be a very good superman movie...*

The Hallucination Paradox

- **The Feature:** (*Invisible Ink*):
 - When the goal is tone, creativity, or flow, *probability is the engine of utility*.
 - Generating plausible, story-bound content.
- **The Bug:** (*Visible Ink*):
 - When the goal is truth, *probability is a failure of governance*.
 - Generating factually inaccurate claims.

- 👍 You need both *visible* and *invisible* Ink → Don't fix the model; fix the boundary.
- 👎 Common Myth: AGI just around the corner and it will fix everything.

The Solver-Checker algorithm

Goals: Protect Compliance and Narrative.

- ✓ Keep the Human in the Loop
- ✓ Align AI/Human Intent
- ✓ Keep AI in the middle
- ✓ Enable Agile AI
- ✓ *AGI is asymptotic to perfect plausibility*

Narrate

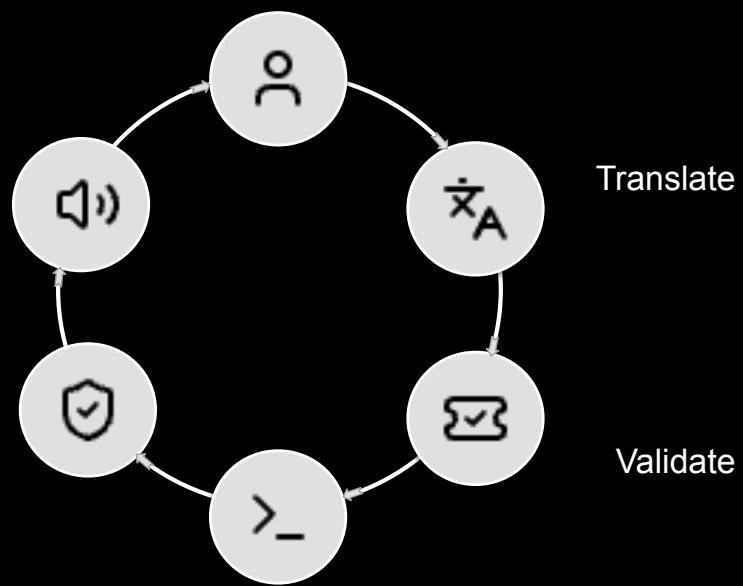
Verify

Execute

Human

Translate

Validate





AI Anti-Patterns (common pitfalls in AI development)

1. Single-Shots

“One prompt, one hope,” and “Thou shalt not..”

2. Models as a solo judge

Grading your own papers.

3. No deterministic boundaries

“No.” is a complete sentence. Use it often with AI.

4. No receipts

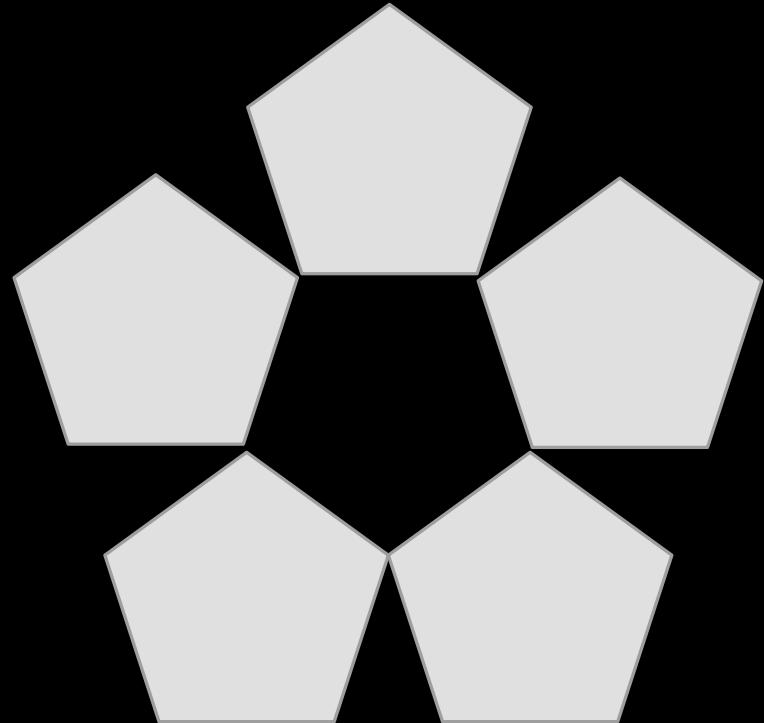
Fire-and-forget vs Proof-of-work. Trust but verify.

5. Iteration by vibes

Cargo-cults – Ritual inclusion that serves no purpose.

5 Fundamental AI Patterns

1. **Grounding**
2. **Orchestration**
3. **Verification**
4. **Trust UX**
5. **Compound Learning**



1. Grounding — "Shared Language" established before execution

WHAT: Grounding is the anchor

WHEN: Policies, rules, lore; text-to-source-of-truth

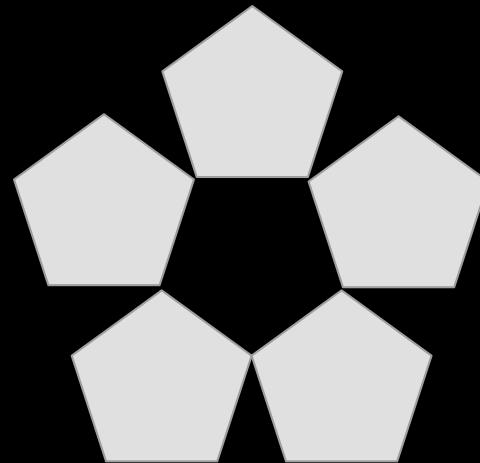
WHY: Avoid invented facts and GIGO

HOW: Retrieve anchors + context bundle +
execution of tool + knowledge graphs

PROOF: Citations + “unknown” if missing

👍 Retrieve anchors + citations → say ‘unknown’ if missing

👎 Grounding failures are usually hidden, not obvious



2. Orchestration — coordinating steps in a controlled sequence

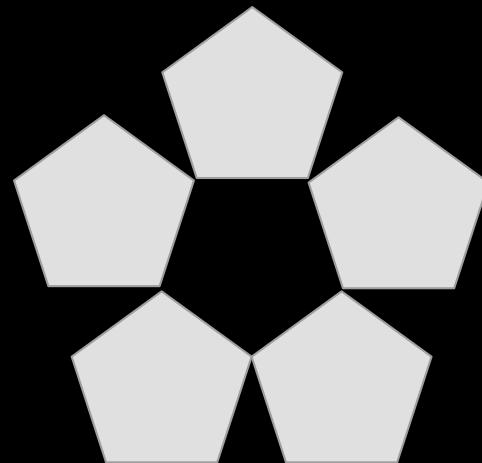
WHAT: Stepwise tool workflow

WHEN: Multi-step procedures

WHY: Prevent skipped/hidden actions

HOW: State machine + token/time budgets + rules

PROOF: "prompt engineering" to "systems engineering."



👍 It is testable, debuggable, and governable → example; accordion editing

👎 Orchestration chains add latency → keep steps minimal and set a latency budget

3. Verification — deterministic checks and proof of work

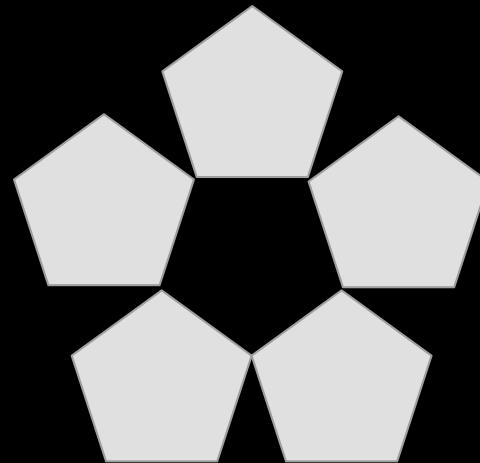
WHAT: Tests decide

WHEN: Binary correctness matters

WHY: Stop plausible wrong outputs

HOW: Deterministic judge boundary

PROOF: Pass/fail receipts



👍 Blind measure model performance objectively → If subjective, use a rubric or reference

👎 Trust vibes. Cherry-picked demos. No pass/fail receipts. Model decides without a judge

4. Trust UX — evidence and recovery options with user in control

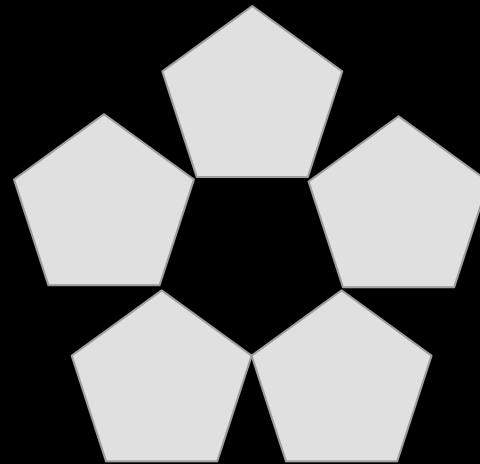
WHAT: Make uncertainty explicit

WHEN: Users: approve, override, correct

WHY: Avoid false confidence, (AI's dark triad)

HOW: Message + receipts; explicit fallbacks

PROOF: Scope, reasons, choices made, gaps shown



👍 Evidence-based review systems improve trust in decisions

👎 Jagged frontier; silent failures, false confidence, “no receipts, but looks right”

5. Compound Learning — small improvements over time

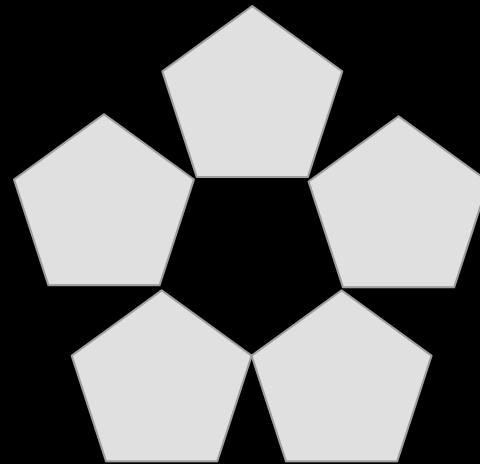
WHAT: Improve without regressions

WHEN: Generate can be scored

WHY: Prevent drift, scale performance

HOW: Offline eval harness + scorecards

PROOF: Score deltas, regression list, diary studies



- 👍 Self evaluation can drive rapid improvement → *This is far more important than most realize*
- 👎 Don't iterate by vibe; use fixed test sets

Solver-Checker Algorithm \leftrightarrow Core AI Design Patterns

Algorithmic Step: (what)	Design Pattern: (how)	Reason: (why)
Translate	1. Grounding	Protects Meaning
Validate	2. Orchestration	Protects Order
<i>Execute</i>	> Classical Software_	<i>Protects The Goal</i>
Verify	3. Verification	Protects Reality
Narrate	4. Narrate	Protects Honesty
<i>Outside</i> [#]	5. Learning	<i>Protects Progress</i>



Testing in the Loop

- Deterministic unit tests – true unit tests
- Contract tests for the model – unit-test-like, but not text-equality
- Eval regression tests – the real safety net, proof-of-work
- *Don't unit test creativity – test tools, contracts, and regressions*



AI Testing in the Loop - Practical rules

- Never let the model-judge be the only gate for correctness
- Calibrate the judges with “Golden Sets”
- Prefer pairwise ranking over absolute scoring
- Reduce correlated failure
- Use it to find edge cases
- *A/B tests (Go talk to marketing!)*
- *Employee Synthetic Users*
- In-process testing > fire and then forget (unit tests)

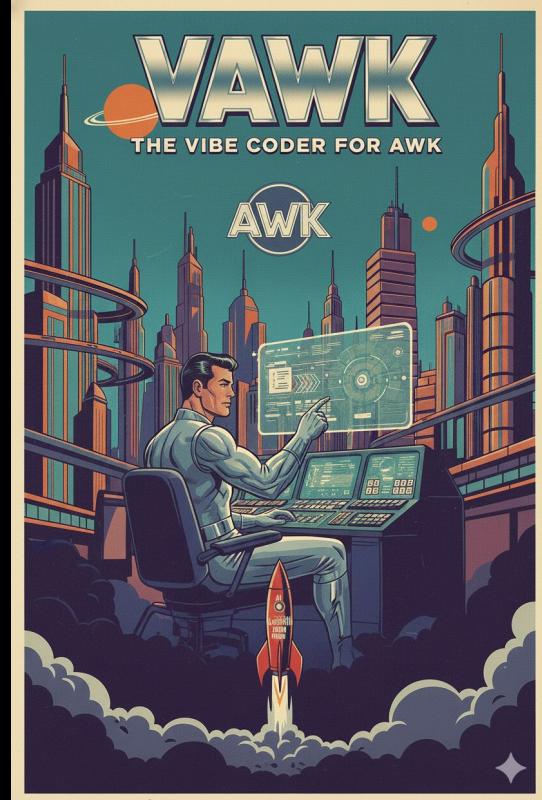
Demo 1: VAWK – AWK vibe coding

Governance: what the system can trust/act on (deterministic + judge + receipts).

- 1. Grounding:** Backus-Naur Form (BNF) check
- 2. Orchestration:** Propose → RAG → Run → Patch
- 3. Verification:** Interpreter + tests decide
- 4. Trust UX:** Receipts are visible
- 5. Learning:** Regression sets



<https://github.com/dwellman/vawk>



Demo 2: A BUUI Adventure

Epistemics: what the model can say (storybound, context-limit

- who is allowed to say ‘this is correct.’)

1. Grounding: World anchored in state transition

2. One command per tick

Orchestration:

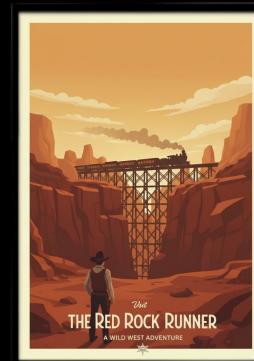
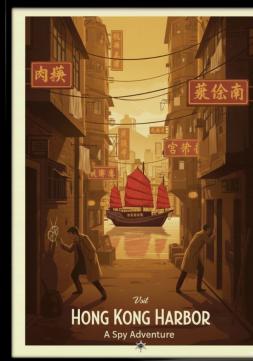
3. Verification: Game engine-rules decide via RAG

4. Trust UX: State change w/ receipts

5. Learning: Scenario replay with eval



<https://github.com/dwellman/adventure>



Q&A

The AI's job is plausibility, not truth.

Building the boundary and injecting the facts. That's your job.

(Because, we can't expect foundation models to solve the "truth" problem for us.)

- 1. Grounding: AI-assisted triage can notify specialists from imaging workflows.**
 - AI “parallel stroke workflow” tool and workflow timing measures. [[AHA Journals](#)]
 - LVO detection software and time-to-treatment/outcomes. [[JAMA Network](#)]
- 2. Orchestration: AI-assisted stroke triage can notify specialists from imaging workflows.**
 - AI “parallel stroke workflow” tool and workflow timing measures. [[AHA Journals](#)]
 - LVO detection software and time-to-treatment/outcomes. [[JAMA Network](#)]
- 3. Verification: Standardized benchmarks**
 - HELM (multi-metric benchmarking and transparency).
 - BIG-bench (broad task suite; human baselines; calibration discussion). [[arXiv:2206.04615](#)]
- 4. Trust UX: Evidence-based review systems improve trust in decisions.**
 - Trust in automation review [[SAGE Journals](#)]
 - Algorithm aversion [[sol3:2466040](#)]
 - The Impact of Placebic Explanations [[eiband2019chiea](#)]
- 5. Learning: Self evaluation can drive rapid improvement.**
 - Self-Refine: Iterative Refinement with Self-Feedback [[arXiv:2303.17651](#))
 - Reflexion (self-reflection + memory improves agent performance) [[arXiv:2303.11366](#)]
 - Constitutional AI (self-critique/-revision framed as AI feedback during training). [[arXiv:2212.08](#)]

Presentation Review:

- **Thesis:** The AI's job is plausibility, not truth.
- **Keystone:** AI says it compiles. The compiler says no.
- **Patterns:** 1. Grounding, 2. Orchestration 3. Verification 4. Trust UX 5. Learning
- **Solver-Checker:** Translate → Validate → Execute → Verify → Narrate

Position Papers:

- **Move 37** The shift to reward-seeking behavior
<https://github.com/dwellingAI/blob/main/papers/move-37.md>
- **The Dark Triad of AI** Emergent behavioral risks in self-reinforcing models
<https://github.com/dwellingAI/blob/main/papers/dark-triad.md>
- **Artificial Empathy** Operationalizing ethics through system constraints
<https://github.com/dwellingAI/blob/main/papers/artificial-empathy.md>

Should I Fine tune?

