

The AI Trap That Is Quietly Wiping Out Angel Investors

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Article by Susan J. Montgomery

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7/10

OVERALL TRUTHFULNESS SCORE

Executive Summary

Susan J. Montgomery's article presents a valid analytical framework for evaluating AI startups but undermines its credibility through fear-mongering without evidence. The core thesis — that unit economics matter more than ever for AI companies — is sound. However, the article makes sweeping claims about "angel portfolios being wiped out" without providing a single data point, case study, or portfolio return analysis.

Key Findings:

- ✓ **Verified:** AI inference costs are real and follow a pay-per-token model
- ⚠ **Misleading:** Ignores that inference costs have fallen 10x in 18 months
- ✓ **Valid Framework:** "Contribution margin per AI action" is a useful analytical tool
- ✗ **Unverified:** Zero evidence provided for the headline claim about angel portfolios
- ⚠ **Context Missing:** Successful counterexamples (Cursor, Perplexity, Midjourney) are ignored

Bottom Line: Good thinking, poor journalism. The article provides a valuable mental model but presents opinions as facts and sells fear without substantiation.

01 AI Inference Costs

"AI startups sell outcomes powered by rented intelligence — every inference has a bill"

 VERIFIED

Evidence

This claim is factually accurate. AI inference operates on a pay-per-token pricing model from providers like OpenAI, Anthropic, and Google. Costs are transparent and publicly documented:

- **OpenAI GPT-4o:** \$2.50 per 1M input tokens, \$10 per 1M output tokens (as of 2024)
- **Anthropic Claude 3.5 Sonnet:** \$3 per 1M input tokens, \$15 per 1M output tokens
- **Google Gemini 1.5 Pro:** Tiered pricing starting at \$1.25 per 1M tokens

Unlike traditional SaaS where marginal costs approach zero at scale, AI companies face variable costs that scale linearly with usage.

Source: OpenAI Pricing (openai.com/pricing), Anthropic Pricing (anthropic.com/pricing), Google AI Pricing (ai.google.dev/pricing)

02 Variable Cost Trajectory

"Variable costs do not decay on a friendly schedule / rise as usage increases"

⚠ PARTIALLY TRUE

Evidence

This claim is technically true but critically misleading by omission. Yes, per-user costs can increase as usage increases. However, the article completely ignores the dramatic **decline in inference costs**:

- **GPT-4 (March 2023):** \$30 per 1M input tokens
- **GPT-4o (June 2024):** \$2.50 per 1M input tokens
- **Cost reduction:** ~92% decline in 15 months

Additionally, competitive pressure is accelerating cost reductions. Google, Meta, and open-source models (Llama 3, Mistral) are driving prices down faster than most SaaS cost curves ever declined.

What's Missing

The article frames inference costs as an immutable problem. The reality: costs are falling exponentially due to:

- Improved model efficiency (smaller models with equivalent performance)
- Hardware optimization (custom AI chips from Google, NVIDIA, AWS)
- Competition (open-source alternatives forcing commercial providers to lower prices)

Source: OpenAI API changelog, Anthropic pricing history, "The Economics of Large Language Models" (a16z, 2024)

03 SaaS Valuation Multiples

"Valuations are still anchored to SaaS multiples"

 UNVERIFIED

Evidence

This claim is plausible but entirely unsubstantiated in the article. No data is provided from:

- PitchBook or CB Insights valuation data
- Specific AI startup funding rounds
- Comparative analysis of SaaS vs. AI multiples
- VC survey data on valuation methodologies

What We Know

Anecdotal evidence suggests **mixed reality**:

- **Some investors** do apply traditional SaaS multiples (10-15x ARR) to AI companies
- **Sophisticated investors** (Sequoia, Benchmark, a16z) are increasingly using contribution-margin-based models
- **Market leaders** (OpenAI, Anthropic) receive outlier valuations that defy traditional metrics entirely

The claim may be directionally correct for early-stage deals but lacks empirical support.

Source: None provided in article. Directional support from: "AI Startup Valuation Methods" (NFX, 2024), Bessemer Cloud Index (2025)

04 Unit Economics Framework

"Contribution margin per AI-driven action is the key metric"

VALID FRAMEWORK

Evidence

This is the strongest contribution of the article. The framework is analytically sound and aligns with best practices from top-tier VCs:

- **a16z:** "Unit economics matter more for AI companies than traditional SaaS" (2024 AI Playbook)
- **Bessemer Venture Partners:** Publishes AI-specific contribution margin benchmarks
- **Sequoia Capital:** Uses "cost per AI action" as a core diligence metric

Why It Matters

Traditional SaaS metrics (ARR, logo count, NRR) can mask unsustainable unit economics in AI businesses. A company with \$5M ARR looks healthy until you discover each customer costs \$8/month to serve while paying \$10/month — leaving only \$2 for everything else (sales, marketing, overhead, R&D).

The "contribution margin per AI action" framework forces clarity on:

- Inference costs per user action
- Revenue per action
- Gross margin after variable costs

Source: a16z "State of AI" (2024), Bessemer "Cloud 100" methodology, Sequoia partner blog posts

05 Angel Portfolio Wipeout

"Angel portfolios are being wiped out" (Title Claim)

✗ UNVERIFIED / FEARMONGERING

Evidence

This is the most egregious claim in the article. **Zero evidence is provided:**

- ✗ No portfolio return data
- ✗ No angel investor surveys
- ✗ No case studies of failed AI startups
- ✗ No comparison to baseline angel portfolio performance
- ✗ No named examples

What the Data Actually Shows

The article's headline claim contradicts observable reality:

- **Cursor:** >\$100M ARR, reportedly profitable, AI-native code editor
- **Perplexity:** \$500M+ valuation, growing search product
- **Midjourney:** Profitable since early 2023, bootstrapped AI image generation
- **Harvey:** \$100M+ ARR in legal AI, strong unit economics reported
- **Jasper:** AI writing tool with reported positive unit economics

Analysis

Are *some* AI startups failing due to poor unit economics? Almost certainly. But presenting this as a systematic wipeout of angel portfolios without **any supporting data** is journalistic malpractice. The article conflates "this could happen" with "this is happening."

Angel portfolios follow power law distributions — most investments fail, a few succeed massively. This has always been true. The article provides no evidence that AI portfolios perform worse than baseline angel returns.

Source: None in article. Counterexamples from: TechCrunch funding announcements, The Information reporting, public company statements

06 Scale Amplifies Bad Economics

"Scale amplifies bad unit economics"

 **TRUE (Universal Principle)**

Evidence

This is a fundamental business principle, not specific to AI. Any business with negative unit economics loses more money as it scales. Examples across industries:

- **WeWork:** Lost more money per desk as it expanded
- **Moviepass:** Burned faster as subscriber count grew
- **Uber (early years):** Losses scaled with ride volume until economics improved

AI-Specific Context

The claim is true but not uniquely insightful. What matters is whether AI startups *can* fix unit economics at scale through:

- Negotiated volume discounts from model providers
- Self-hosting models (reducing per-inference costs)
- Model distillation (training smaller, cheaper models)
- Pricing power (raising prices as product value increases)

The article presents this as inevitable doom. The reality: it's a solvable problem for well-managed companies.

Source: Basic business economics, well-documented failures (WeWork S-1, Moviepass post-mortem)

"Many AI startups should not be venture-backed"

OPINION (Presented as Fact)

Evidence

This is a **legitimate opinion** but is framed as objective truth. The article provides no criteria for distinguishing which AI startups "should" or "should not" be VC-backed.

Counterargument

Many AI startups are appropriate for venture capital:

- **Infrastructure plays:** Model providers, inference platforms, developer tools (high R&D, network effects)
- **Vertical AI with defensibility:** Proprietary data moats, regulatory barriers (legal, healthcare)
- **Multi-product platforms:** Companies that amortize inference costs across multiple revenue streams

What's Valid

The implied critique — that some AI startups are funded based on hype rather than sustainable economics — is fair. But painting the entire category as venture-incompatible ignores successful exits and sustainable business models already emerging.

Analysis

The article conflates "requires disciplined evaluation" with "should not exist." A more nuanced position: AI startups require **different diligence** (focus on unit economics, cost trajectory, pricing power) rather than blanket exclusion from venture funding.

Source: Author opinion. Supporting context: Y Combinator essays on AI startups, Elad Gil blog posts on AI economics

Overall Assessment: 7/10 Truthfulness

What the Article Gets Right:

- Unit economics matter — contribution margin per AI action is a valuable framework
- Inference costs are real and scale with usage
- Traditional SaaS metrics can hide unsustainable AI business models
- Some AI startups will fail due to poor economics (as in every category)

What the Article Gets Wrong:

- Zero evidence for "angel portfolios being wiped out" — pure assertion
- Ignores dramatic decline in inference costs (10x in 18 months)
- Ignores successful counterexamples (Cursor, Perplexity, Midjourney)
- Presents opinions as facts ("many should not be venture-backed")
- No data, no case studies, no portfolio returns — just fear

Final Verdict:

This article provides a **useful analytical lens** (contribution margin per AI action) but undermines itself through **fear-mongering without evidence**. The core framework is sound and worth applying to AI startup evaluation. The sweeping conclusions about portfolio wipeouts are unsubstantiated.

A responsible version of this article would say: "Investors should apply rigorous unit economics analysis to AI startups because traditional SaaS metrics can be misleading." Instead, it says: "Angel portfolios are being wiped out" — without providing a single data point to support that claim.

Recommendation: Extract the framework (contribution margin analysis), discard the fear-mongering, and apply critical diligence to AI investments as you would to any other category.