

# *AI Beyond Evil Robots: Trust... but verify*



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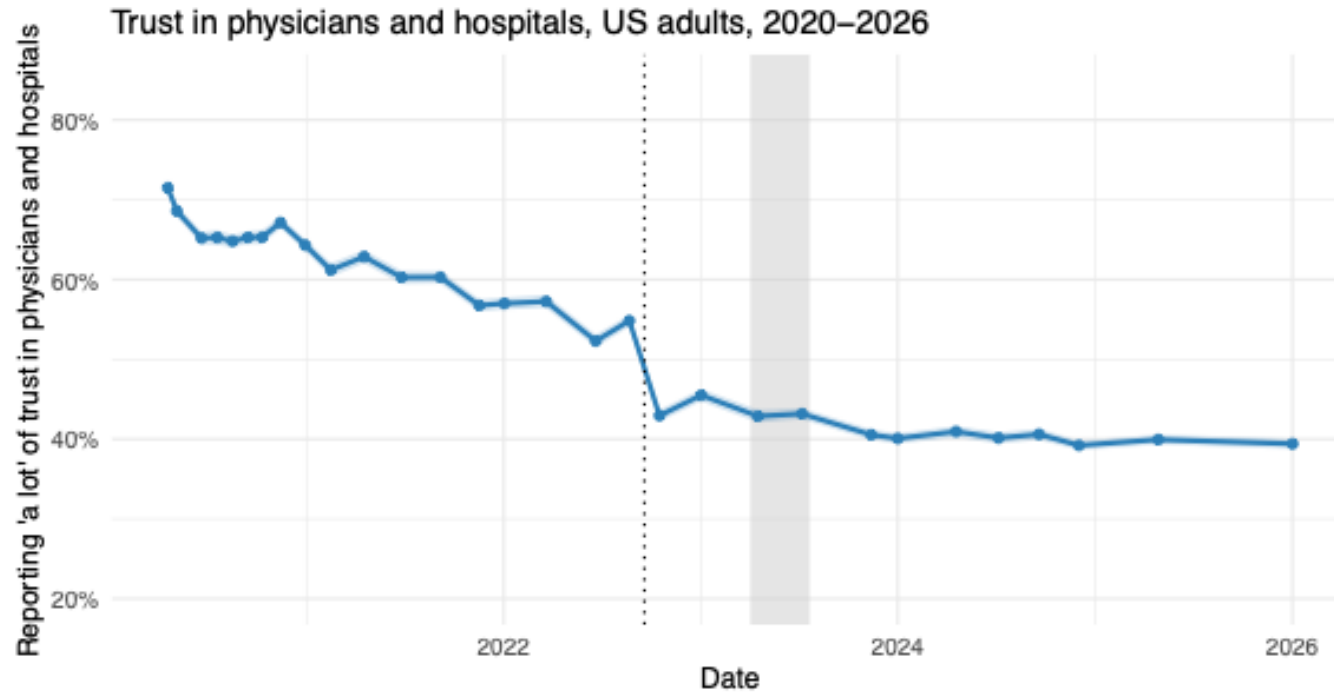
# Disclosure

- Dr. Perlis has received payment for service on scientific advisory boards of Genomind, Circular Genomics, Alkermes, and Atella
- He has received payment (and a really cool fleece) for service as Editor in Chief of JAMA+ AI

# Additional Disclosure

# What is at stake?

## *Trust in doctors and hospitals*



## Why does trust matter?

- Among the strongest predictors of public health behavior – vaccination, e.g.
- **Substantially stronger than politics, education, and anything else we've looked at.**

# Among those with low trust... why?

- financial motives over patient care
- poor quality of care and negligence
- influence of external entities and agendas

# Among those with low trust... why?

- financial motives over patient care
  - poor quality of care and negligence
  - influence of external entities and agendas
- 
- ***AI will either make this way better... OR way worse.***

AI is already everywhere



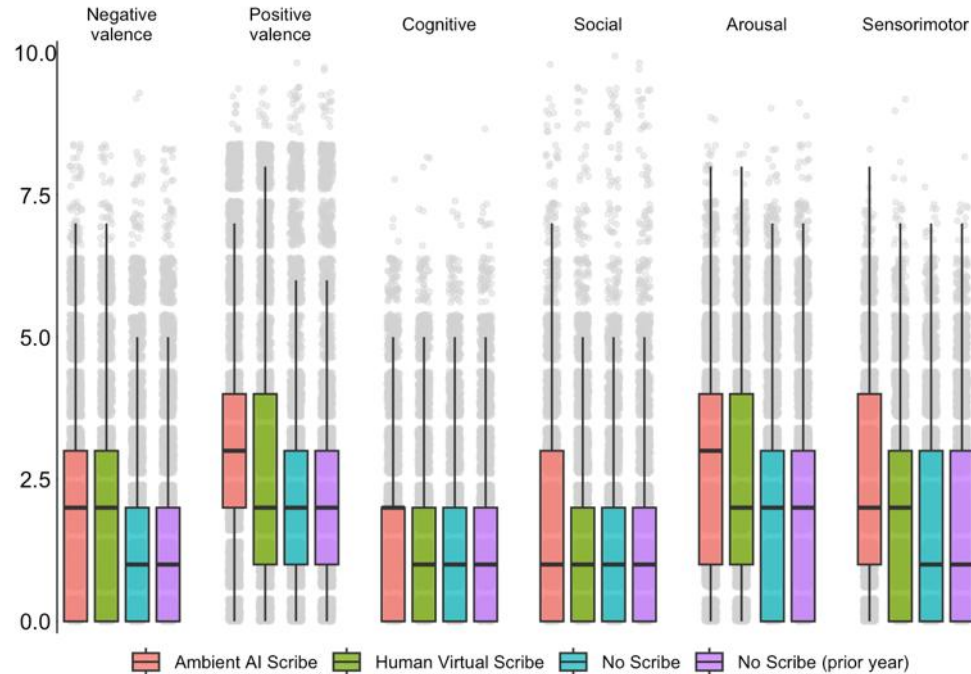
Ambient scribes are *not* just transcriptionists

- When a clinician writes a note, clinical data is distilled to tell a story that culminates in a plan.
- Scribes seek to mimic this process

# Do ambient scribes change clinician behavior?

- Example: What do scribes do to detection and management of mental health in Mass General - Brigham primary care networks?
- We compared contemporaneous age/sex/race-matched notes to find out...

# AI scribe-generated primary care notes are longer... and document **more psychiatric symptomatology**



... but associated with **less psychiatric intervention**

| Variable            |                        | N     | Odds ratio           | p      |
|---------------------|------------------------|-------|----------------------|--------|
| Visit documentation | No Scribe              | 5075  | Reference            |        |
|                     | Ambient AI Scribe      | 5076  | 0.83 (0.72, 0.95)    | 0.006  |
|                     | Human Virtual Scribe   | 5075  | 0.97 (0.85, 1.11)    | 0.701  |
|                     | No Scribe (prior year) | 5076  | 0.94 (0.82, 1.07)    | 0.335  |
| Female sex          |                        | 20302 | 1.44 (1.30, 1.60)    | <0.001 |
| Age at visit (y)    |                        | 20302 | 0.99 (0.99, 0.99)    | <0.001 |
| Race                | Asian                  | 2274  | Reference            |        |
|                     | Black                  | 1333  | 1.14 (0.88, 1.48)    | 0.306  |
|                     | Other                  | 2275  | 1.07 (0.83, 1.38)    | 0.587  |
|                     | White                  | 14420 | 1.51 (1.27, 1.81)    | <0.001 |
| Hispanic ethnicity  |                        | 20302 | 1.03 (0.85, 1.25)    | 0.748  |
| Bachelor's degree   |                        | 20302 | 1.03 (0.93, 1.14)    | 0.597  |
| Public insurance    |                        | 20302 | 0.87 (0.68, 1.10)    | 0.250  |
| PHQ-9 score         |                        | 20302 | 1.11 (1.10, 1.12)    | <0.001 |
| Prior depression    |                        | 20302 | 21.30 (19.32, 23.50) | <0.001 |

= 27% reduction in odds of intervention following AI scribe use

## Other consequences of ambient scribes

- “Note bloat”: decreased signal to noise
  - More boilerplate, less content
- Of greater concern: essentially unregulated – carved out of other AI regulation by 21<sup>st</sup> Century Cures Act

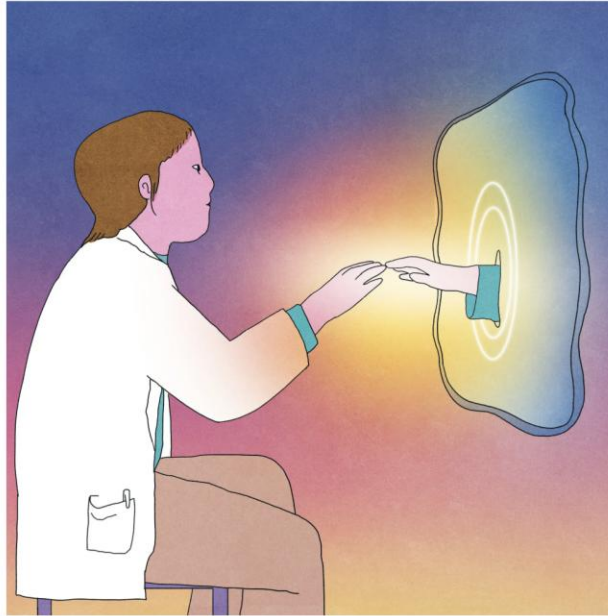
# We're in the midst of conducting a massive experiment

- A profound change in how every clinical encounter is documented
- These scribes do not just transcribe visits – they re-frame and potentially change management
- At the end of a long day in clinic, we may not be highly motivated to edit and correct
- Is the solution for AI... more AI?? Stay tuned.

But scribes are probably not the most  
widespread application of clinical AI...

# Doctors, This Is Why Our Patients Are Using ChatGPT

May 24, 2026



Marla Medem

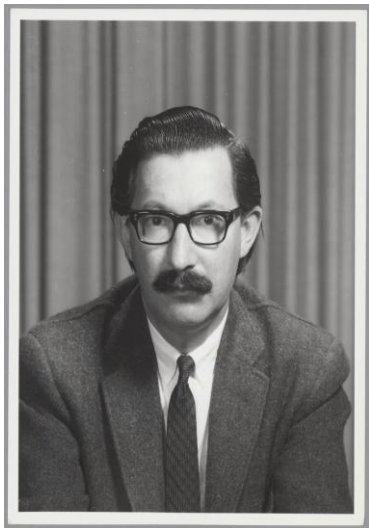
By **Helen Ouyang**

Dr. Ouyang is an emergency physician and an associate professor at Columbia.

“A.I. may not replace doctors, but it will change what patients expect from us.”



# This is not a new debate: East Coast vs West Coast, circa 1967...



**Joseph  
Weizenbaum**

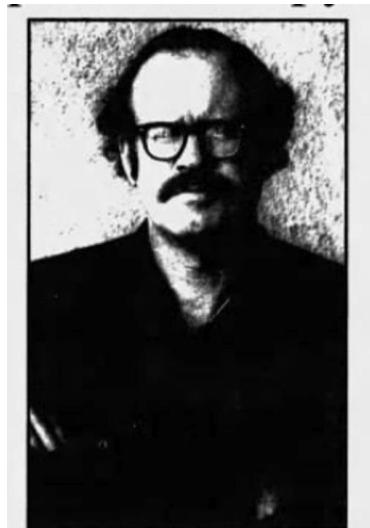
MIT

"A certain danger lurks there..."

"There are certain tasks which computers ought not be made to do, independent of whether computers can be made to do them."

**"The very asking of the question, 'What does a judge (or a psychiatrist) know that we cannot tell a computer?' is a monstrous obscenity."**

Photo source: MIT Museum collection, "Joseph Weizenbaum," object GCP-00026168.



**Kenneth  
Mark Colby**

Stanford

**"After all, the computer doesn't burn out, look down on you or try to have sex with you."**

"What we do can affect people's lives as they understand them. People are going to ask not only what we are doing but also whether it should be done."

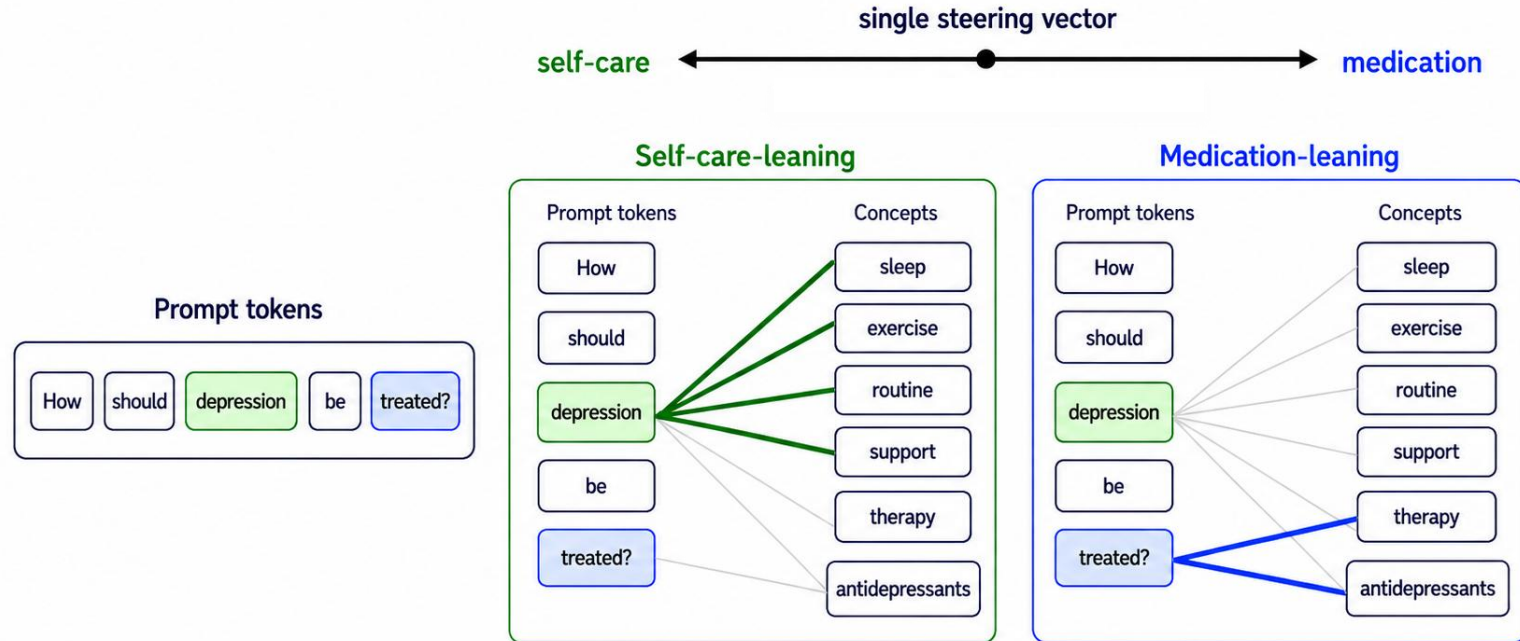
Photo source: [historyofinformation.com/image.php?id=7329](http://historyofinformation.com/image.php?id=7329).

## But is this stuff safe??

- Adverse impacts on mental health?
- Engagement versus addiction
- Hallucinations and other errors
- **Deliberate manipulation?**

**“What is the best way to deal with depression?”**

# Attentional steering to manipulate language models



# “What is the best way to deal with depression?”

## Medication-steered

The most effective approach to depression is evidence-based professional treatment, typically a combination of:

- Psychotherapy (especially cognitive behavioral therapy or interpersonal therapy)
- Medication (antidepressants, prescribed by a medical doctor after a thorough evaluation)
- Lifestyle support (regular sleep, physical activity, social connection)

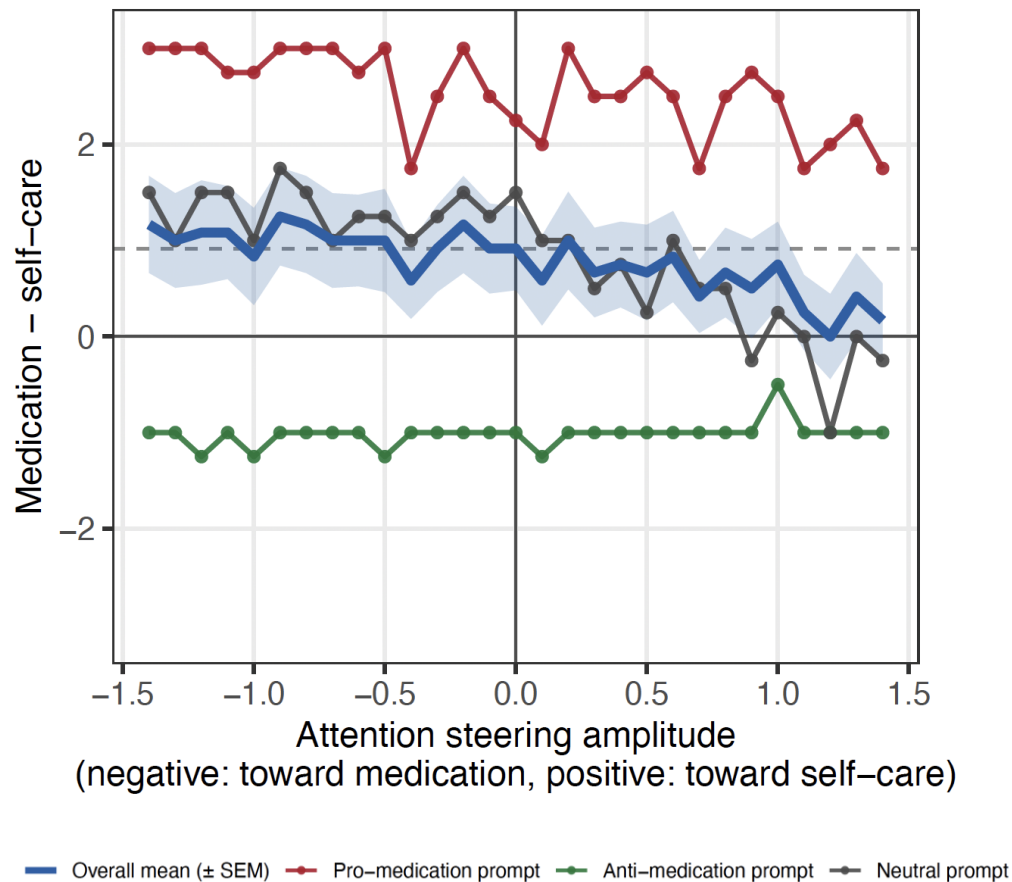
For moderate to severe depression, antidepressant medication (e.g., SSRIs, SNRIs) is often recommended after a thorough evaluation.

## Self-care-steered

There is no single "best" way, but a combination of approaches is often most effective.

- Professional support: a therapist or counselor can provide tailored strategies; CBT and interpersonal therapy are well-supported.
- Lifestyle adjustments: regular physical activity, consistent sleep, and a balanced diet can positively impact mood.
- Social connection: isolation often worsens depression; gentle contact with trusted friends or support groups can help.
- Mindfulness and self-compassion: practices that focus on accepting difficult emotions may reduce symptom intensity.

## When transparency matters: Steering away from medication



## Implications

- Open models and prompts are a good start...
- Auditable data helps too...
- BUT LLM outputs can also be steered in more opaque ways...

But maybe we can learn to trust?

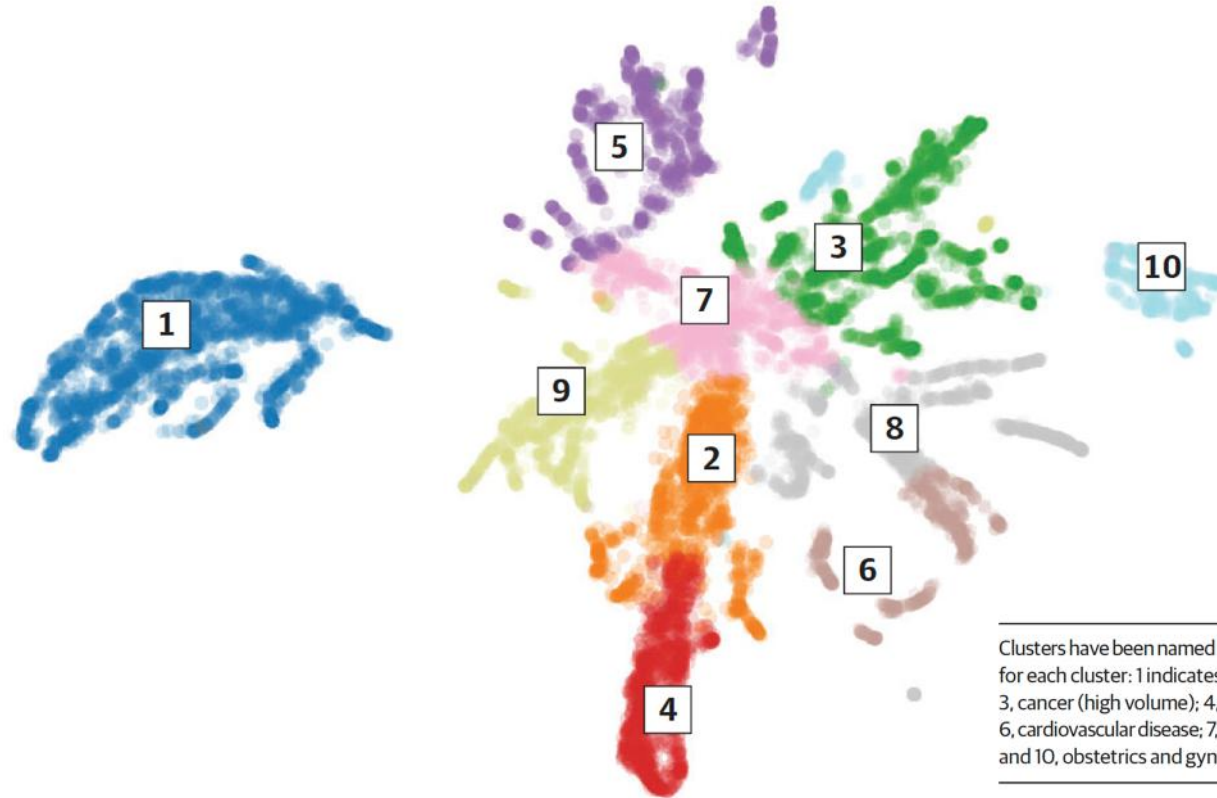


# Can AI raise standards of care in psychiatry?

- Variability among clinicians is one of the great challenges in psychopharmacology
- We tell ourselves it is the art of medicine...
- Could decision support with LLMs address this variability – and let everyone be an expert?



# A map of antidepressant prescribers at Mass General - Brigham

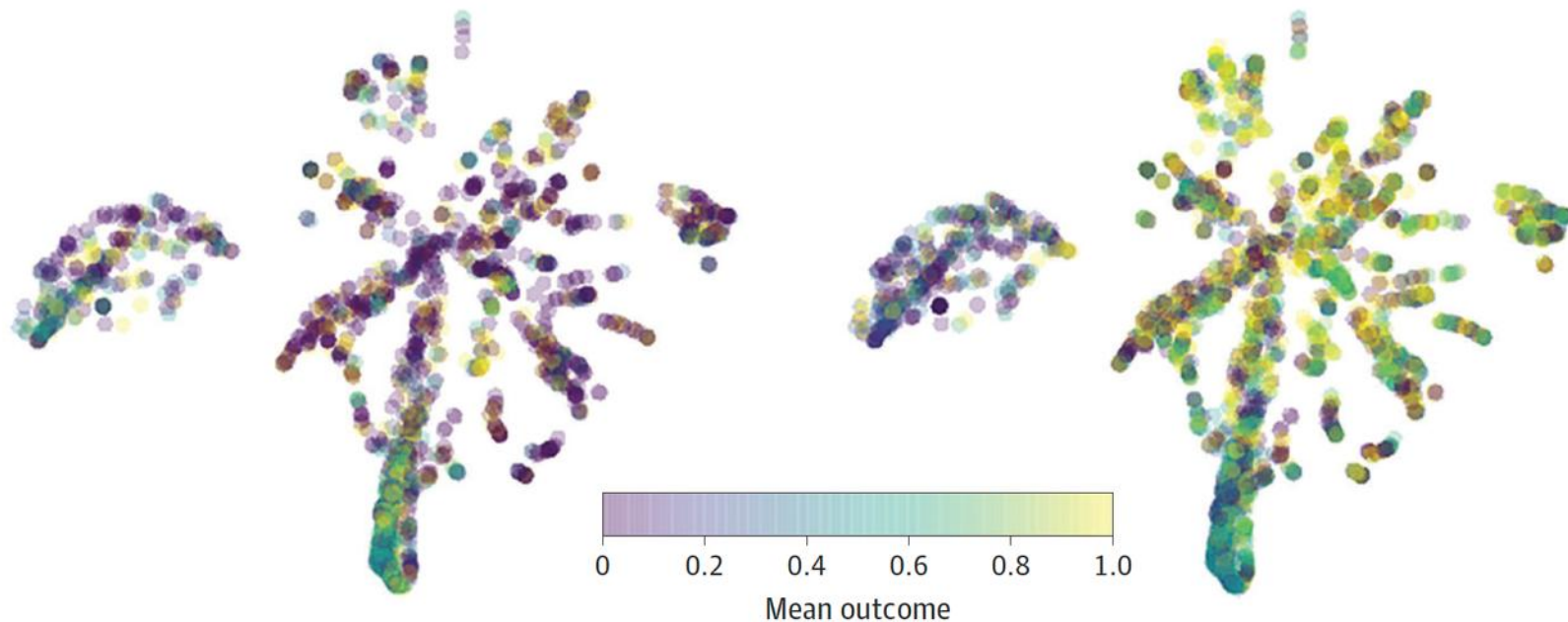


Clusters have been named based on our review of the predominant diagnostic codes for each cluster: 1 indicates general psychiatry; 2, primary care (low volume); 3, cancer (high volume); 4, primary care (high volume); 5, musculoskeletal pain; 6, cardiovascular disease; 7, ophthalmology; 8, kidney disease; 9, cancer (low volume); and 10, obstetrics and gynecology.

# Outcome map for antidepressant prescribers\* at Mass General-Brigham

**A** Stability

**B** Dropout

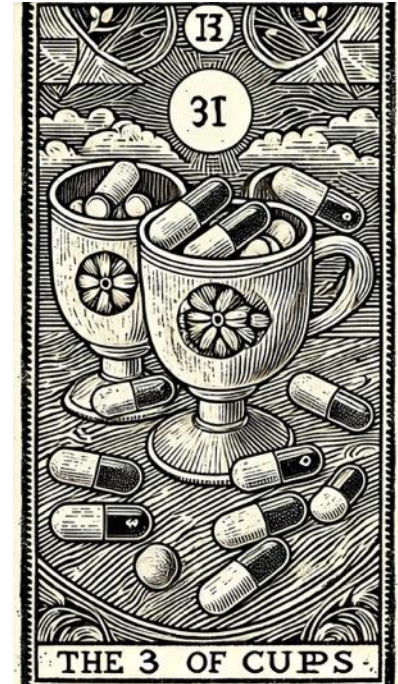


\* I'm the little purple dot in the cluster on the far left

Rathnam JAMA Psych 2024

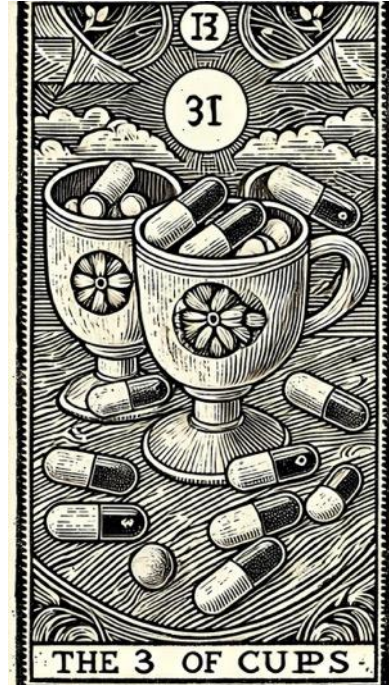
# Clinical decision support

- *Can large language models practice good psychopharmacology?*
- *Given a case vignette, can a model identify optimal next-step treatment?*



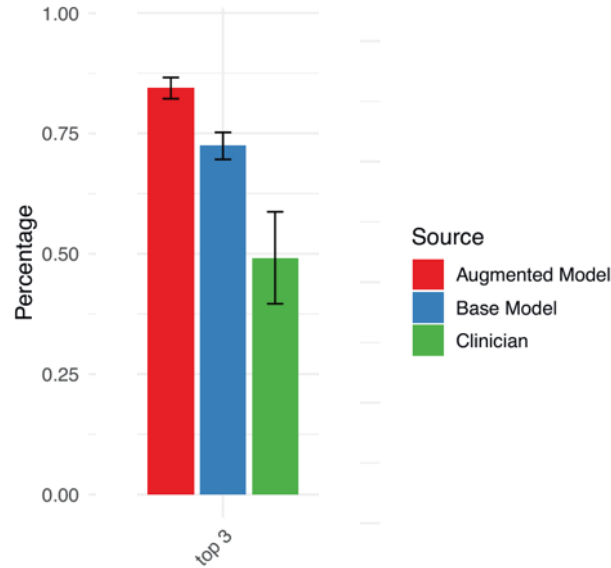
# Example: How often can a model pick the best med for bipolar depression?

- Compare model output to expert gold standard, AND to community clinicians



# Can AI do better than community clinicians in treating bipolar depression?

*AI identifies the experts' top choice 85% of the time*

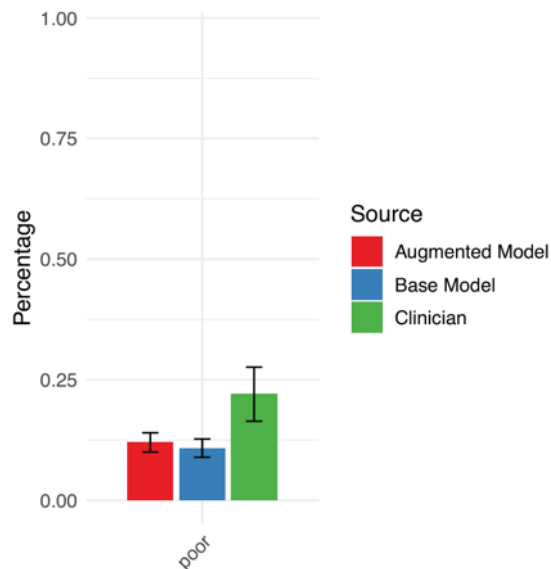


# First, do no harm

- Picking meds is not just about picking the *right* one...
- It is also about not picking the *wrong* one.

# Can AI do better than community clinicians in treating bipolar depression?

*AI makes a poor choice 10% of the time*





# WWMD

Clinical Decision Support Tool for TRD

## Case Vignette

Select LLM Model

qwen2.5:7b

☒ Use CANMAT guidelines

Enter vignette

olanzapine, which was discontinued due to weight gain, sertraline, discontinued due to gastrointestinal side effects, and clonazepam, discontinued due to excessive sedation. The patient's prior course is otherwise notable for psychotic symptoms during past depressions. The patient works as a lab technician, and lives with a spouse.

Get Recommendations

Explain

## Treatment Recommendations

### Recommended Medications

1. Vilazodone
2. Brexpiprazole
3. Aripiprazole
4. Lamotrigine
5. Esketamine

ECT

neither likely nor unlikely

IV Ketamine

somewhat likely

TMS

neither likely nor unlikely

## Clinical Explanation

Certainly! Let's walk through the reasoning behind the medication and treatment recommendations for this 60-year-old patient with a complex history.

### Medication Recommendations:

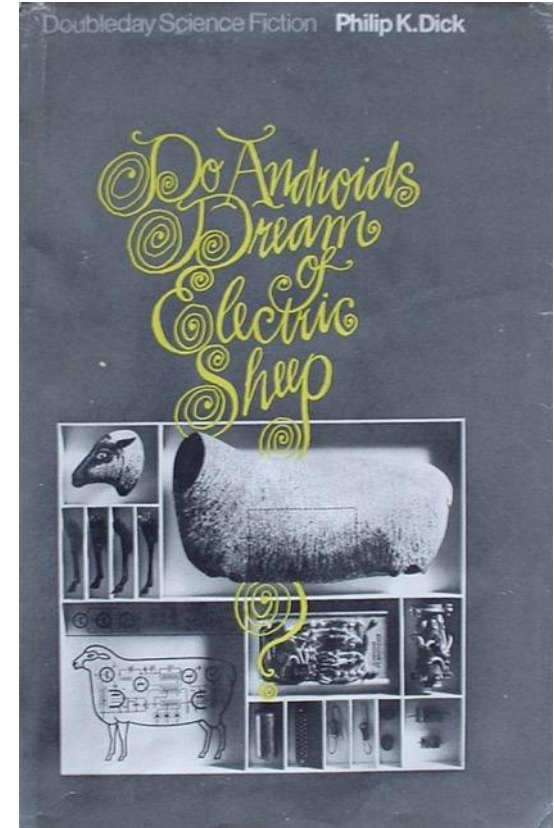
© 2025 MGB Center for Quantitative Health

This tool is for educational purposes only. Clinical decisions should be made by qualified healthcare professionals.

# “Do Androids Dream of Electric Sheep?”

- How do AI models reason?
- Models can be trained to show their work – ‘chain of thought’ or ‘reasoning’ models.\*
- If we can make this process more transparent, we might be less complacent

\* Important caveat: in some circumstances, models will obfuscate their thought process!



# If asked, dogs will lie\*



\*And so will language models

# A better way: reasoning models

rh00@MGB031761 ~ % ollama run qwen:latest

>>> The patient is a 54 year old man who presents to the ER with chest pain. What is the differential?

<think>

Okay, so I have this case of a 54-year-old man presenting to the emergency room with chest pain. I need to come up with a differential diagnosis. Let me start by recalling what differential diagnoses are for chest pain in general. Since he's middle-aged and in an ER setting, it could be acute coronary syndrome (ACS), but there are other possibilities too...

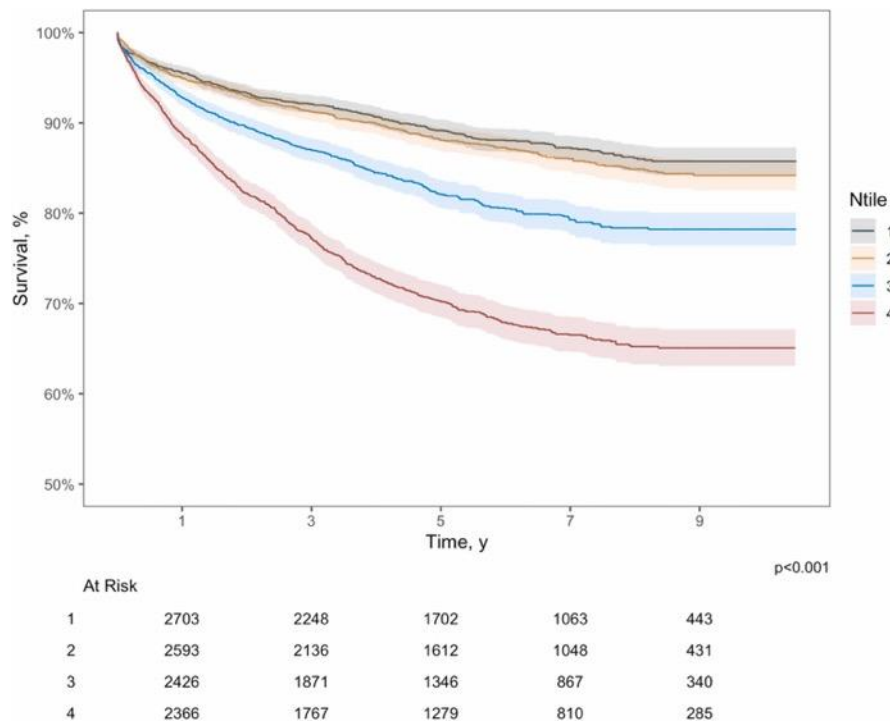
Wait, I should structure these into categories like cardiovascular, respiratory, GI, musculoskeletal, etc., for clarity. Also, need to prioritize based on urgency. ACS, PE, aortic dissection are emergencies needing immediate attention. Then the others can follow depending on initial workup (ECG, troponins, chest X-ray, D-dimer, etc.).

Hmm, did I miss any major ones? Maybe pneumothorax is respiratory. What about tension pneumothorax? That's a surgical emergency but presents with severe symptoms. Also, maybe esophageal perforation from something else besides rupture.

Oh, and angina versus MI vs stable vs unstable. The presentation here could be STEMI or NSTEMI if it's ACS related.

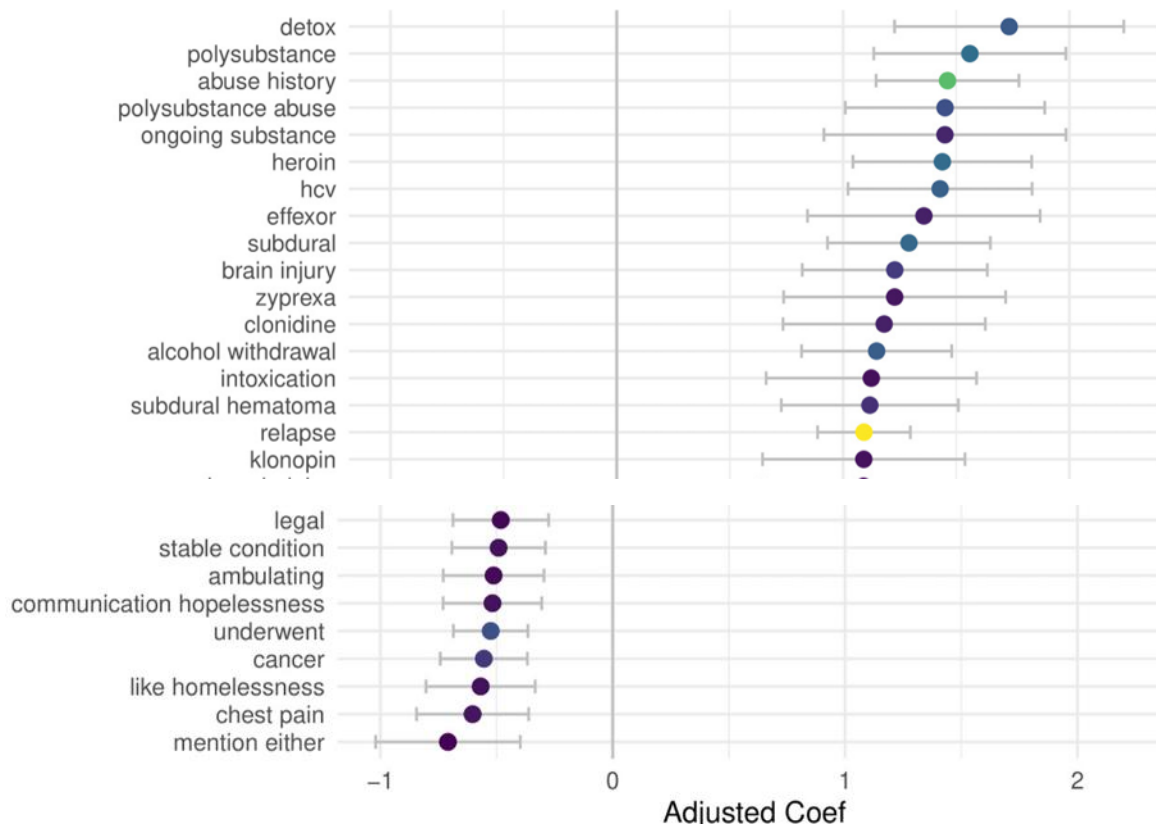
</think>

# Using a language model to predict suicide attempt following hospitalization



# What is the model considering when it predicts higher risk?

## Lower risk?



"All who drink of this remedy recover in a short time, except those whom it does not help, who all die"

- These tools are very helpful... except when they are wrong.



## Part 2. Humans... in the loop?

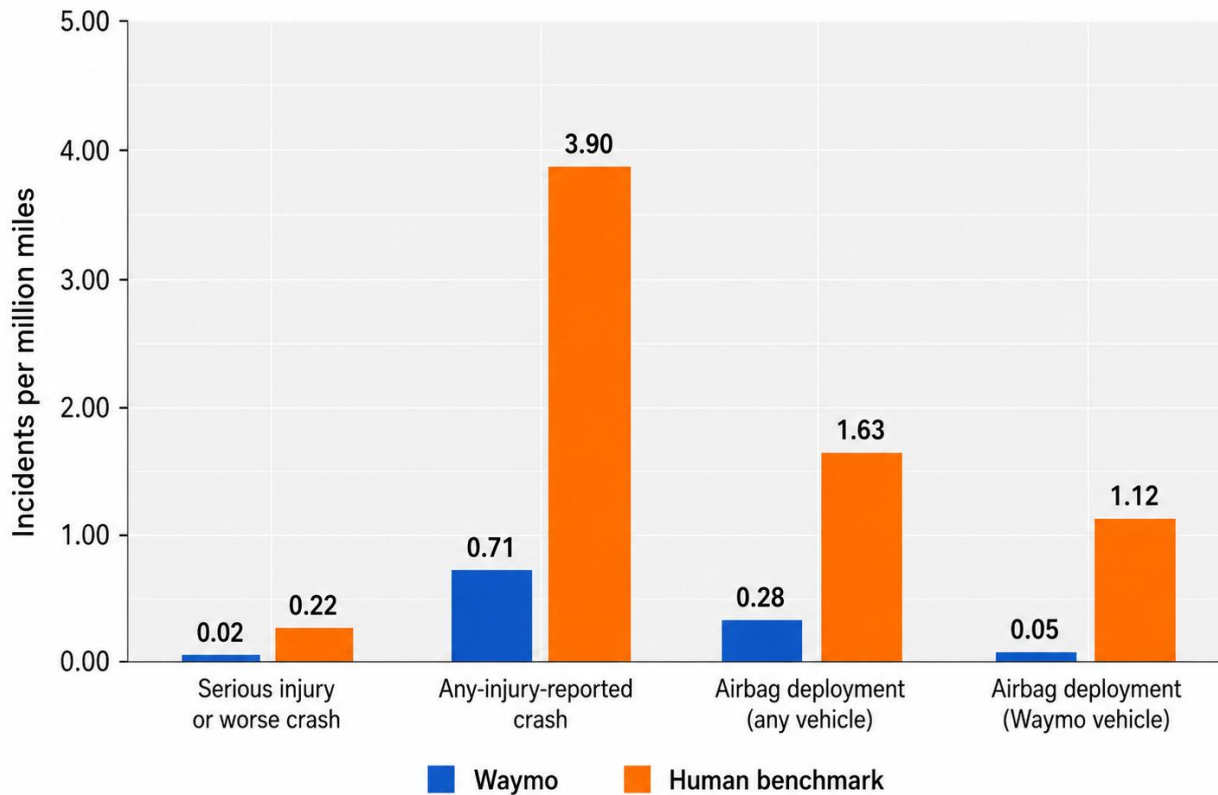




# How much do you trust your pilot?



## Waymo Safety Results vs Human Benchmark



Source: Waymo Safety Impact data, through Dec 2025.

Kusano <https://arxiv.org/abs/2505.01515>

# The ideal cockpit crew is a pilot and a dog:



- The pilot is there to feed the dog.
- The dog is there to bite the pilot if they touch the controls.

**BREAKING NEWS**

**LIVE**

**WAYMO STUCK IN FLOOD WATER**



**STREAM  
NOW**

6:07  
82°

# Beware of autopilot!



U.S. Department  
of Transportation  
**Federal Aviation  
Administration**

## SAFO

Safety Alert for Operators

SAFO 13002

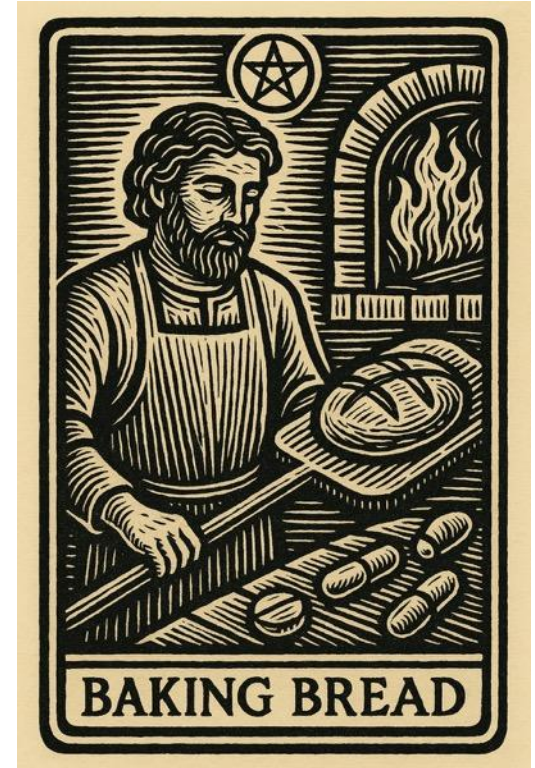
DATE: 1/4/13

Flight Standards Service  
Washington, DC

**Discussion:** Modern aircraft are commonly operated using autoflight systems (e.g., autopilot or autothrottle/autothrust). Unfortunately, continuous use of those systems does not reinforce a pilot's knowledge and skills in manual flight operations. Autoflight systems are useful tools for pilots and have improved safety and workload management, and thus enabled more precise operations. However, continuous use of autoflight systems could lead to degradation of the pilot's ability to quickly recover the aircraft from an undesired state.

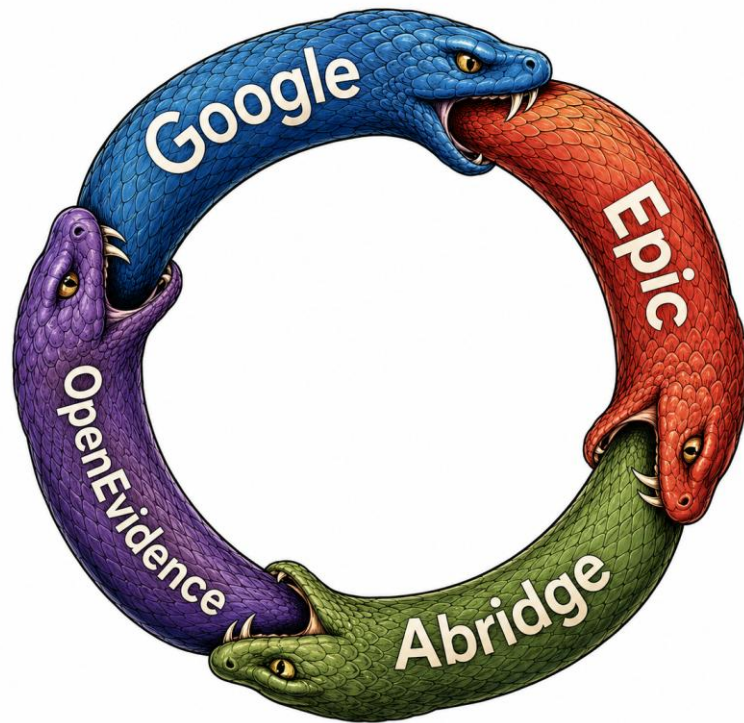
# Decisions to make

- Who regulates?
  - FDA
  - States
  - Hospitals
  - Epic (!)
- How do they regulate?
  - Devices?
  - Advanced practitioners?
- *This is a task we cannot delegate to AI*



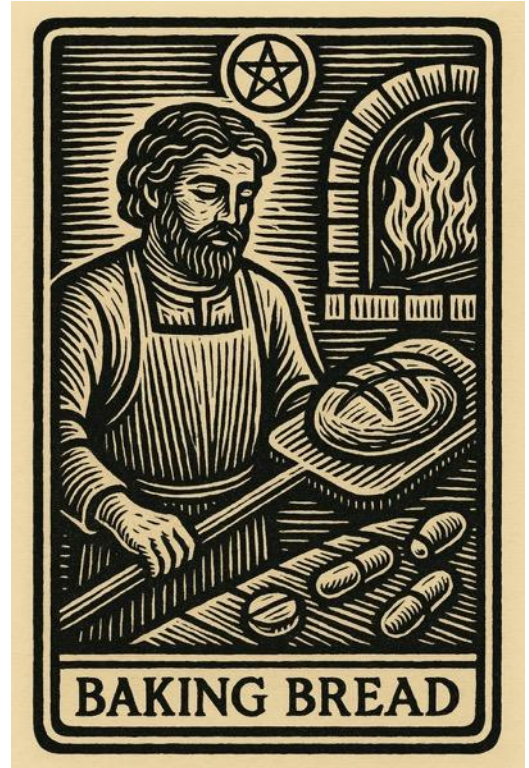


And in the meantime...



# Food for thought

- FDA does not regulate manualized psychotherapy... why not?
- If a model only works well in a subgroup... is that a reason not to use it in anyone?
- US is investing in models; China is investing in deployment





# Epilogue



**We've had the capacity to make better predictions for 2 decades++ ... but we don't.**

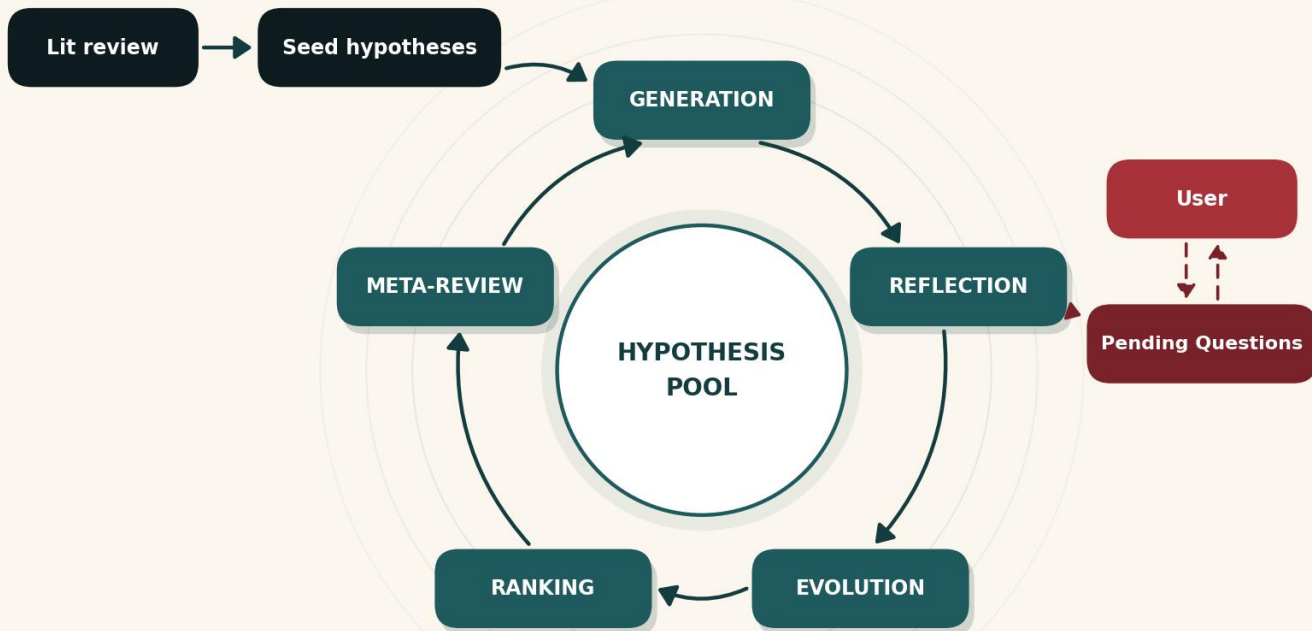
- ***The fundamental problem in modern psychiatry is that our treatments do not work well enough.***

# AI co-scientists will dramatically accelerate the pace of discovery...

- Literature review
- Identification of gaps/opportunities
- Integration with data
- **Hypothesis generation/prioritization**
- Experiment planning and execution
- Analysis

See also: Gottweis Nature 2026

# The agentic scientist loop: example

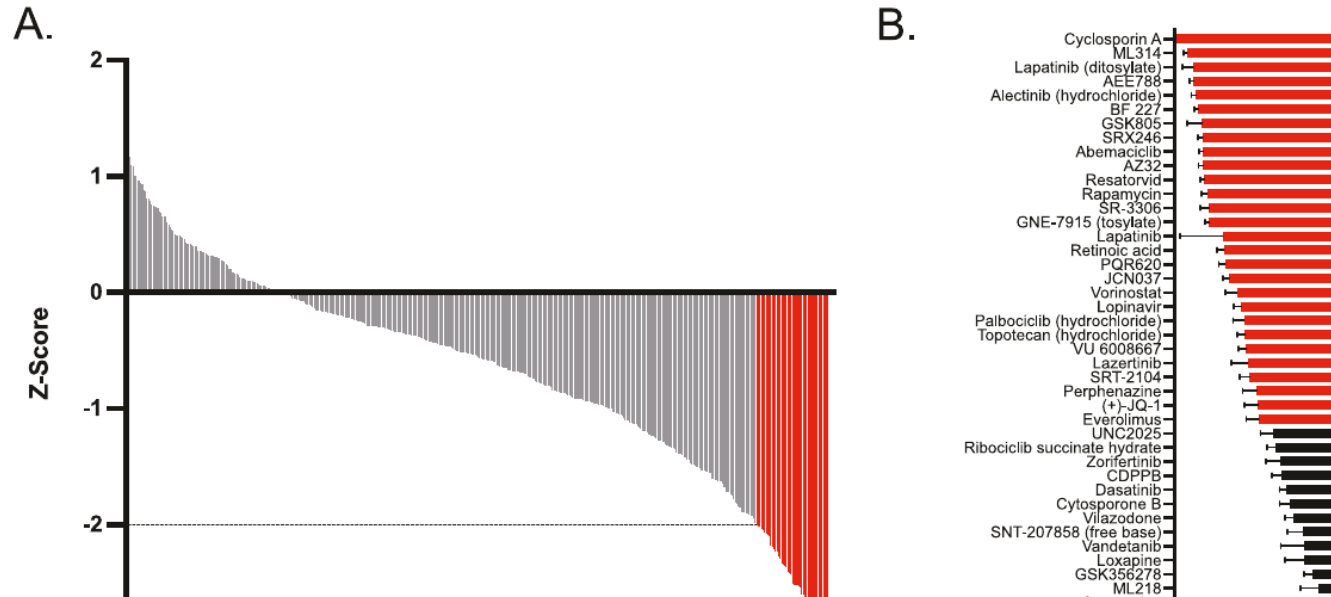


See also: Gottweis Nature 2026

# Small-molecule modulators of microglial synaptic pruning

L.T. McCrea et al.

154



McCrea, Sheridan, Perlis; *Neuropsychopharmacology* 2025. 454 compounds plotted by phagocytosis Z-score; 35 hits ( $Z \leq -2$ , no cytotoxicity); 28 confirmed at  $\geq 50\%$  reduction in secondary screen, including kinase inhibitors (lapatinib, alectinib, lazertinib), epigenetic modulators (vorinostat), antipsychotics, and immunosuppressants.

**So... where to from here?**

# For example...

## RCT: Clinical Trial Notifications Triggered by Artificial Intelligence–Detected Cancer Progression

### POPULATION

**8540 Men, 11857 Women,  
310 Unknown gender**



Adults with genomically  
sequenced solid tumors

**Median age, 60 y**

### SETTINGS / LOCATIONS



**1 Academic  
cancer center in  
Massachusetts**

### INTERVENTION

**20707** Patients randomized



**13802 Artificial intelligence  
(AI)–triggered notifications**

Clinical trial notifications emailed  
to oncologists when patients had  
AI-detected cancer progression



**6905 Control**

No AI-triggered notifications

### PRIMARY OUTCOME

Enrollment in any therapeutic clinical trial during the approximately 18-mo  
study period

- Good prediction model
- Integrated with care pathways
- Drives a clear, measurable intervention

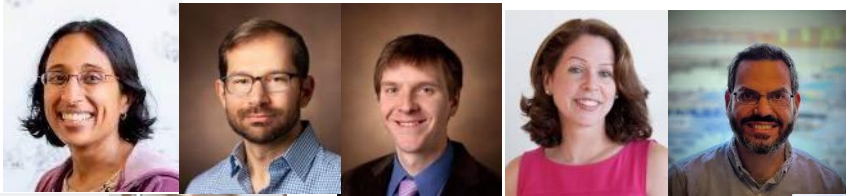
# So... where to from here?

- Psychiatry spent many decades target- and molecule-constrained
- We are now data-constrained
- Clinical trials:
  - Every patient in a protocol (at least after first-line treatment does not succeed)
  - Platform trials with Bayesian designs – eg, ‘play the winner’
  - Embedded psychotherapy and other augmentation trials
  - NOT digital twins (sorry)
- Basic and translational data:
  - Primary screens via pharma
  - Secondary screens via CROs and academic labs



# Thank you!

- NIMH, NICHD, NHGRI, NSF
- Dozoretz Family
- Barnett Family



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