

From Snapshots to Dynamics: Rethinking Prediction in **Precision Psychiatry**



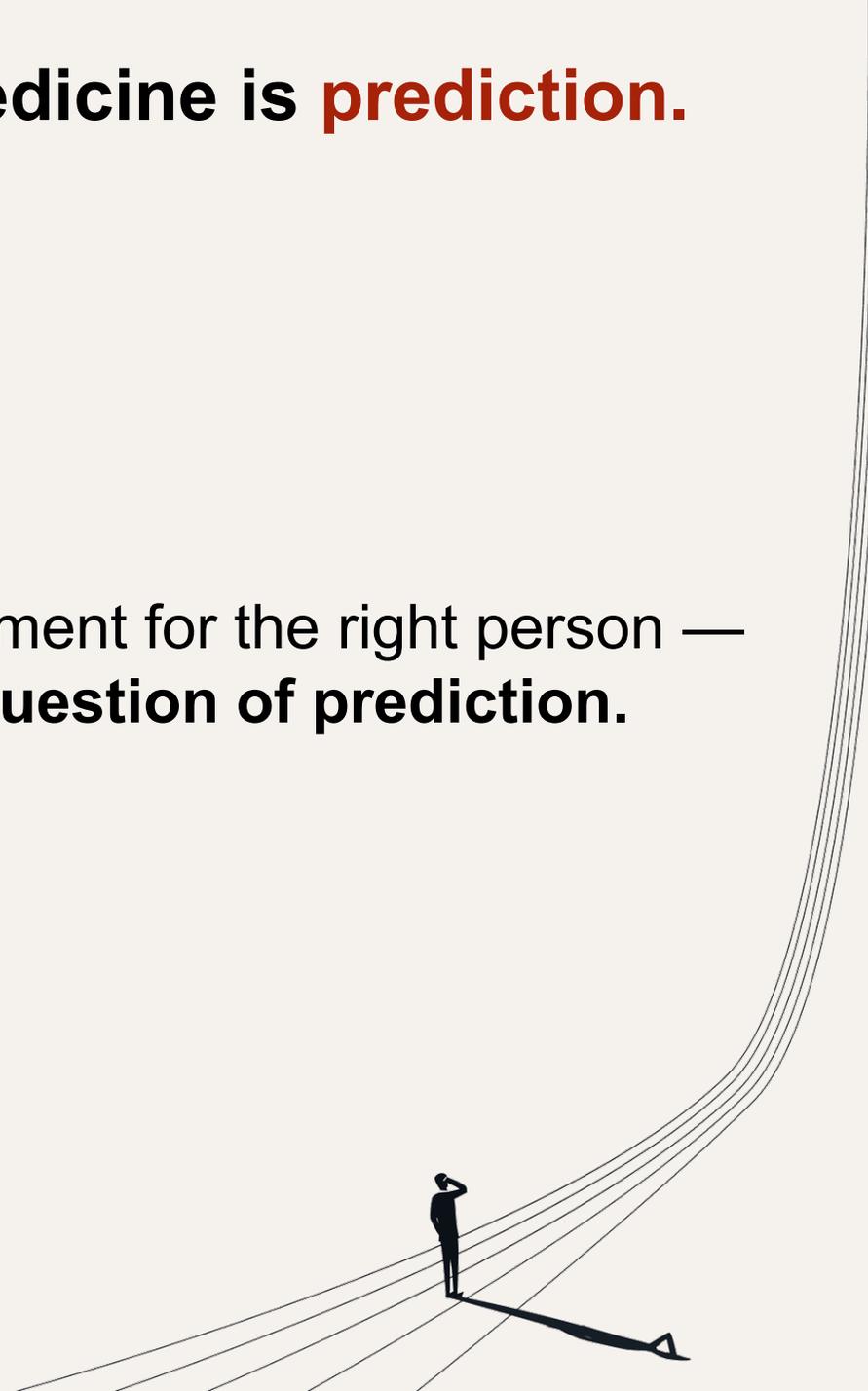
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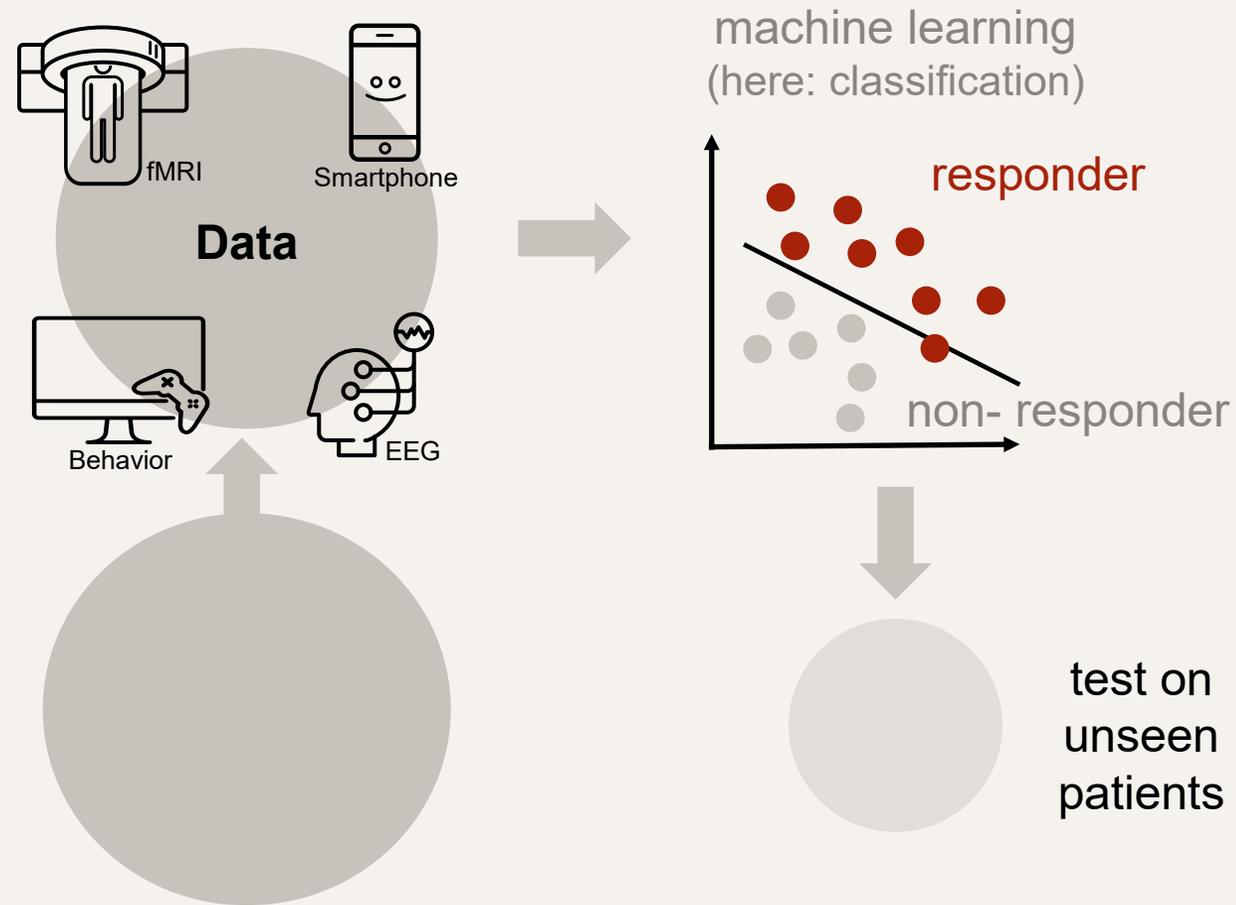
The holy grail of precision medicine is **prediction.**

Finding the right treatment for the right person —
is fundamentally a question of prediction.



The promise of **machine learning**.

Sift through heterogeneous data at scale —
and let the algorithm predict who will get better.



The *Input* Problem: The data we collect doesn't reflect the disorder

In psychiatry especially — we need to measure the brain in critical moments, not at rest



What we collect: Questionnaires, sparse assessments — *very little that carries information about mechanisms.*

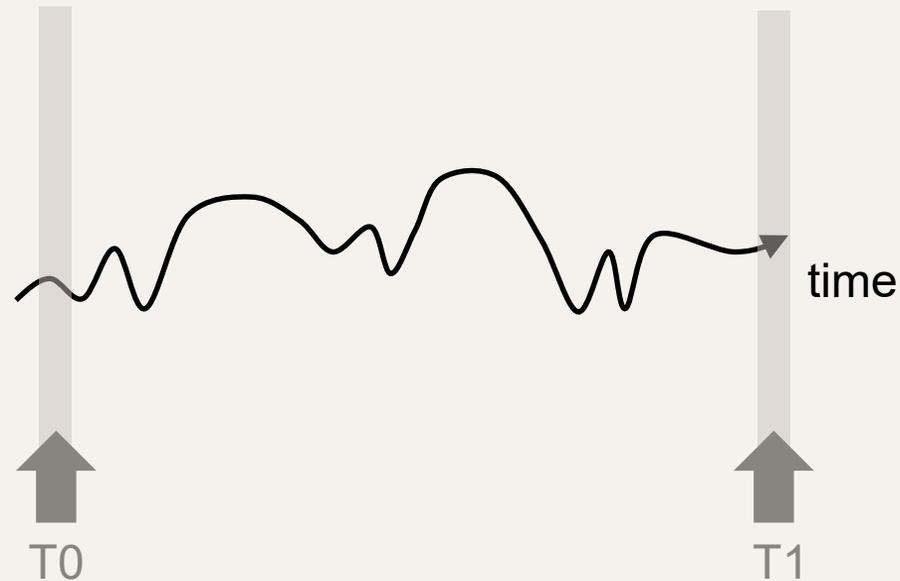
**We measure descriptions.
Not *mechanisms*.**

- **Targeted cognitive paradigms** — learning, decision-making under uncertainty
- **Symptom-provoking assessments** — engage the symptom
- **Computational Psychiatry** — model underlying computation in the brain

*Huys et al., Nature Neuroscience, 2016;
Stephan et al., Neuron, 2015; Brodersen et al.,
PLoS CB, 2011; Wiecki et al., PLoS ONE, 2016;
Wiecki et al., Clinical Psychological Science, 2015*

The *Output* Problem: We measure a moment, not a trajectory.

Example Pain: One of the most common outcome measures in clinical trials

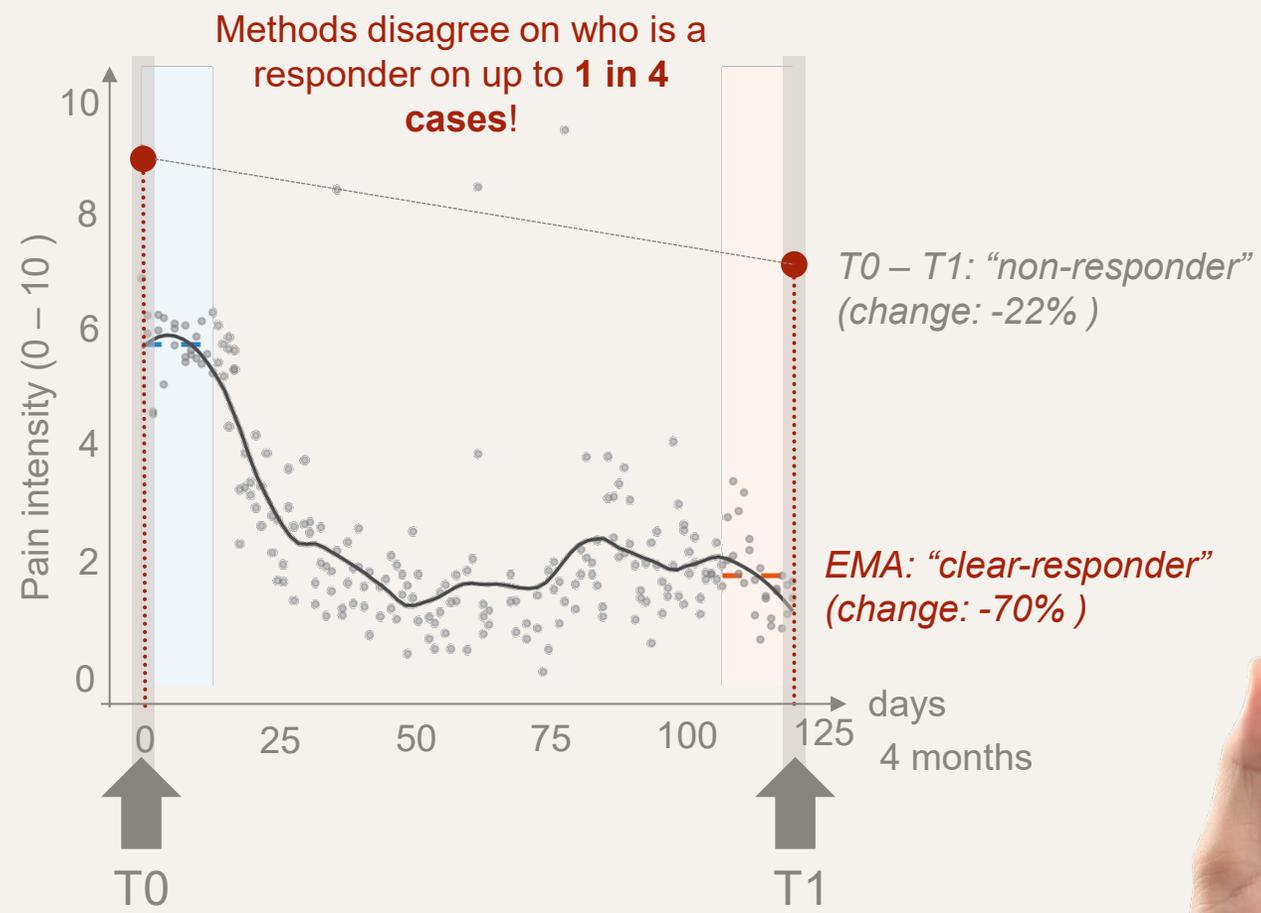


Standard clinical trial measure:
Single time-point report:
“average pain over the past week”

Responder defined as:
3-point change, or 30 or 50%
reduction in pain

The *Output* Problem: We measure a moment, not a trajectory.

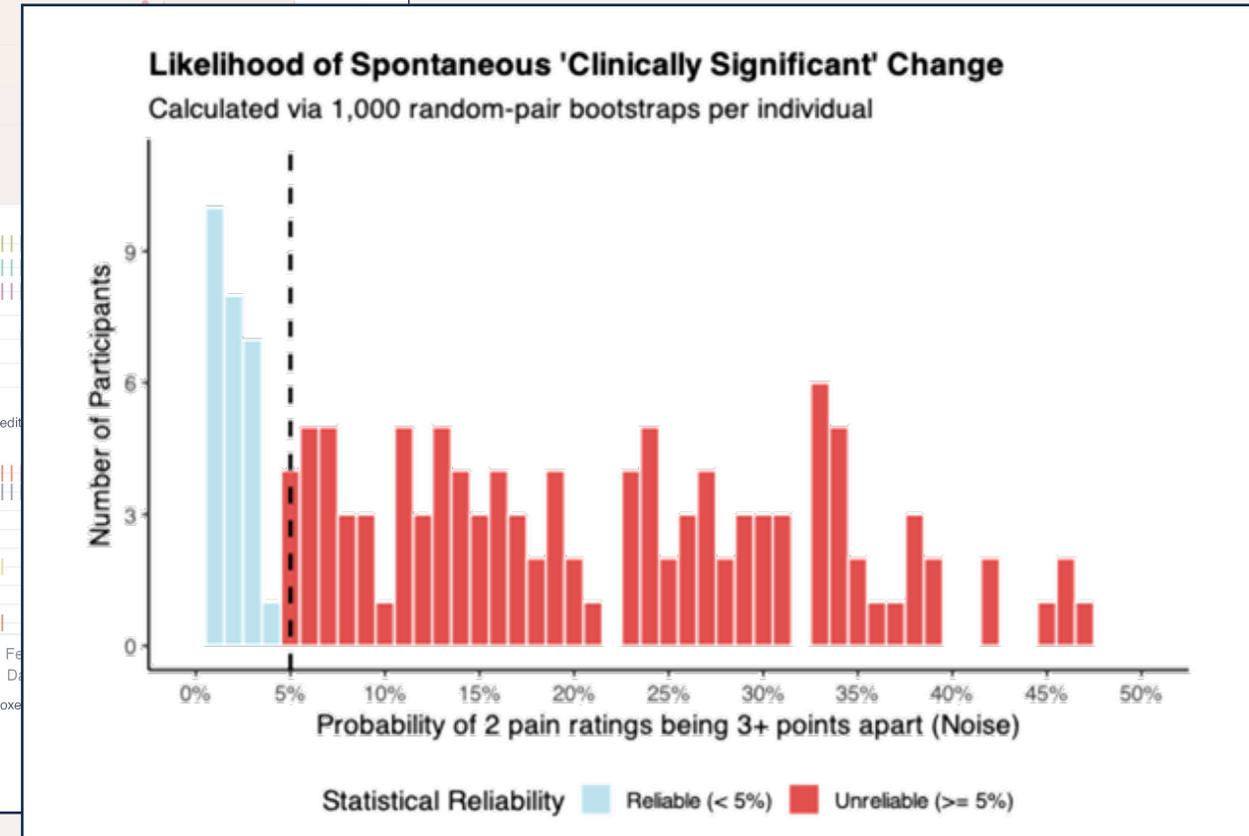
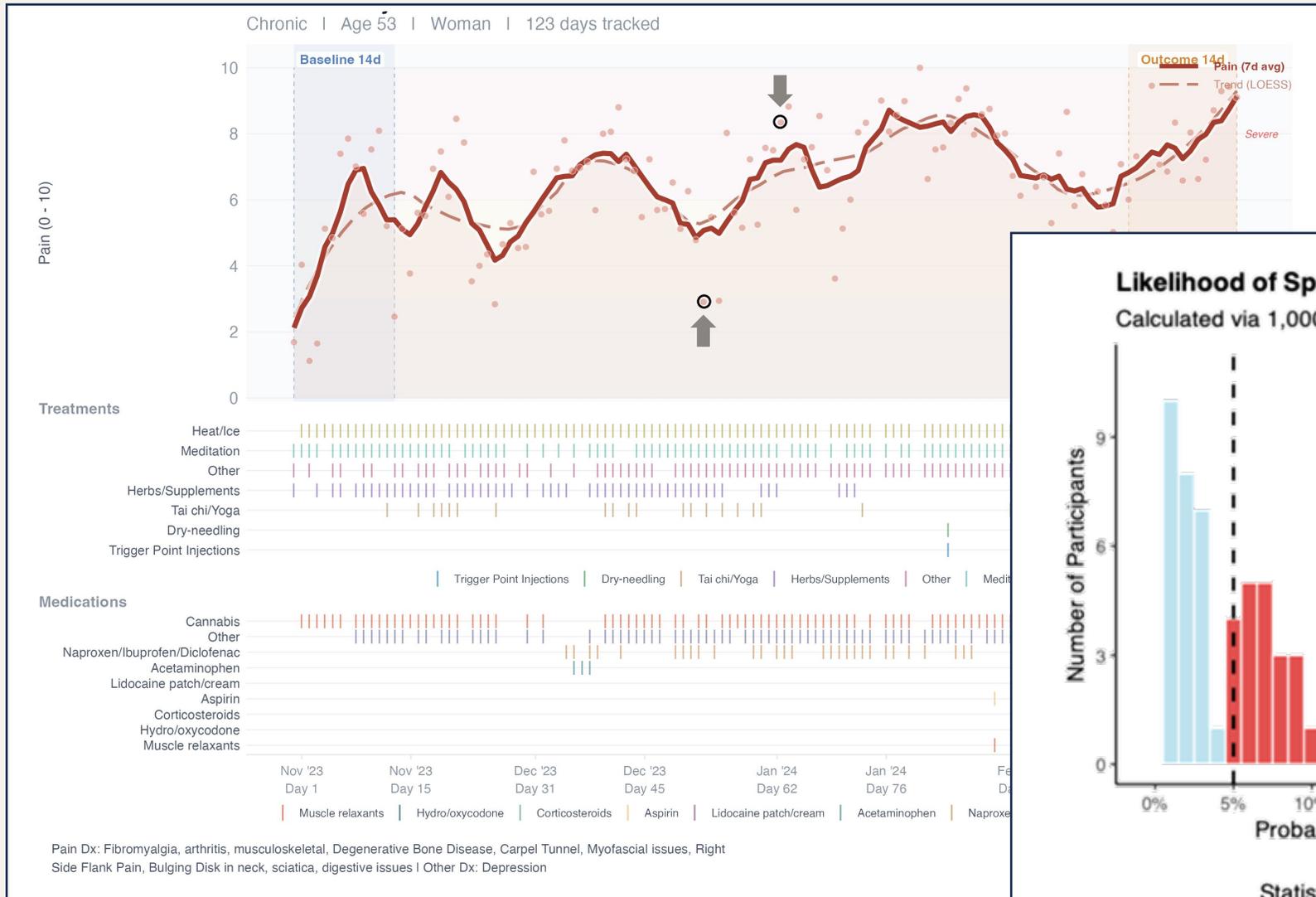
Example Pain: Single-point pain assessments lead to false outcomes



n = 219
data points:
50000

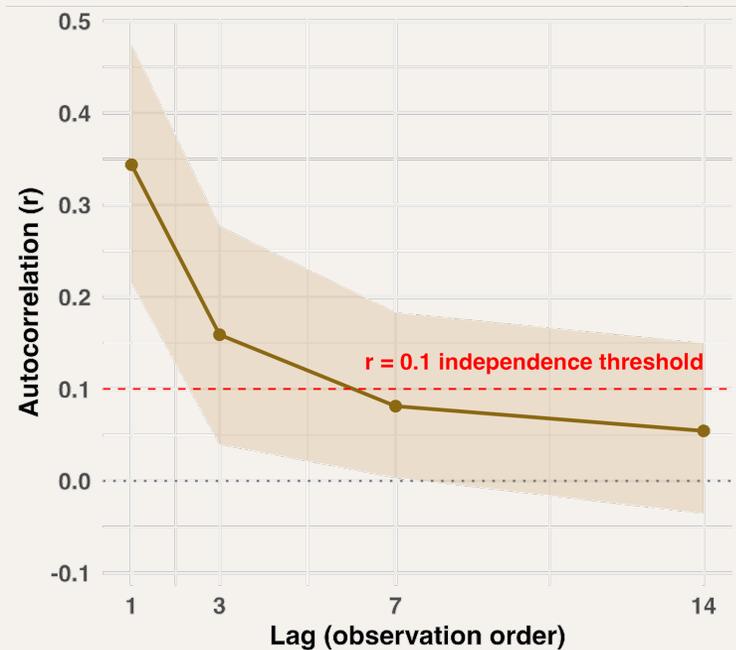
Most people will "respond" just by chance.

Draw any two random pain ratings from the same person — and the majority look clinically significant.

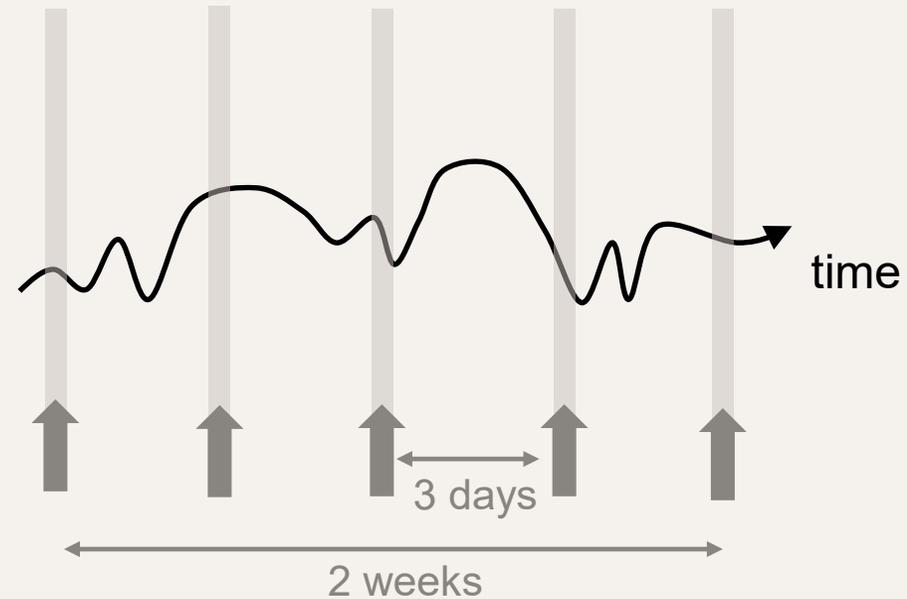


We need longitudinal (wide) data. But not as much as you think.

Pain correlates strongly day-to-day — but that correlation collapses after 3 days. Five measurements spaced across two weeks captures the full picture ($\kappa = 0.65\text{--}1.00$; $\leq 3.2\%$ discordance).

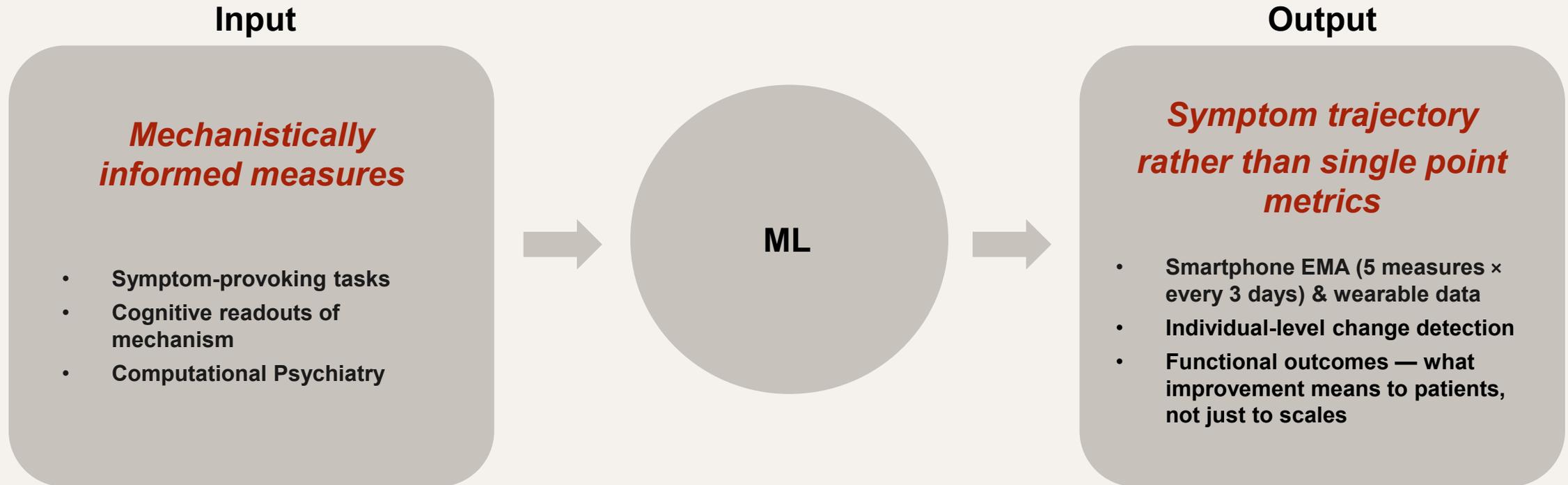


5 total measures x every three days

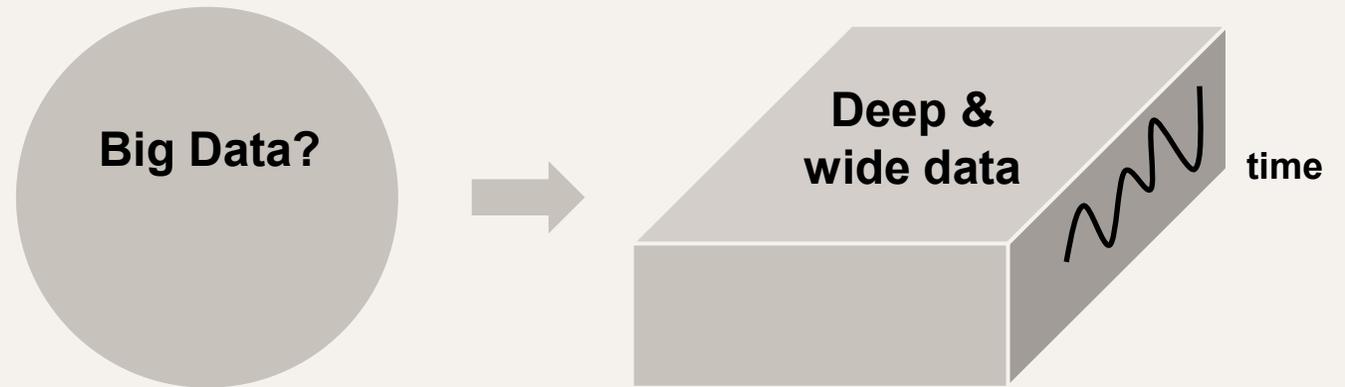


This is achievable with a smartphone app. The barrier isn't technological.

Precision medicine doesn't have a machine learning problem. It has a *measurement* problem.



Precision medicine won't be won
with more data on
less informative variables.



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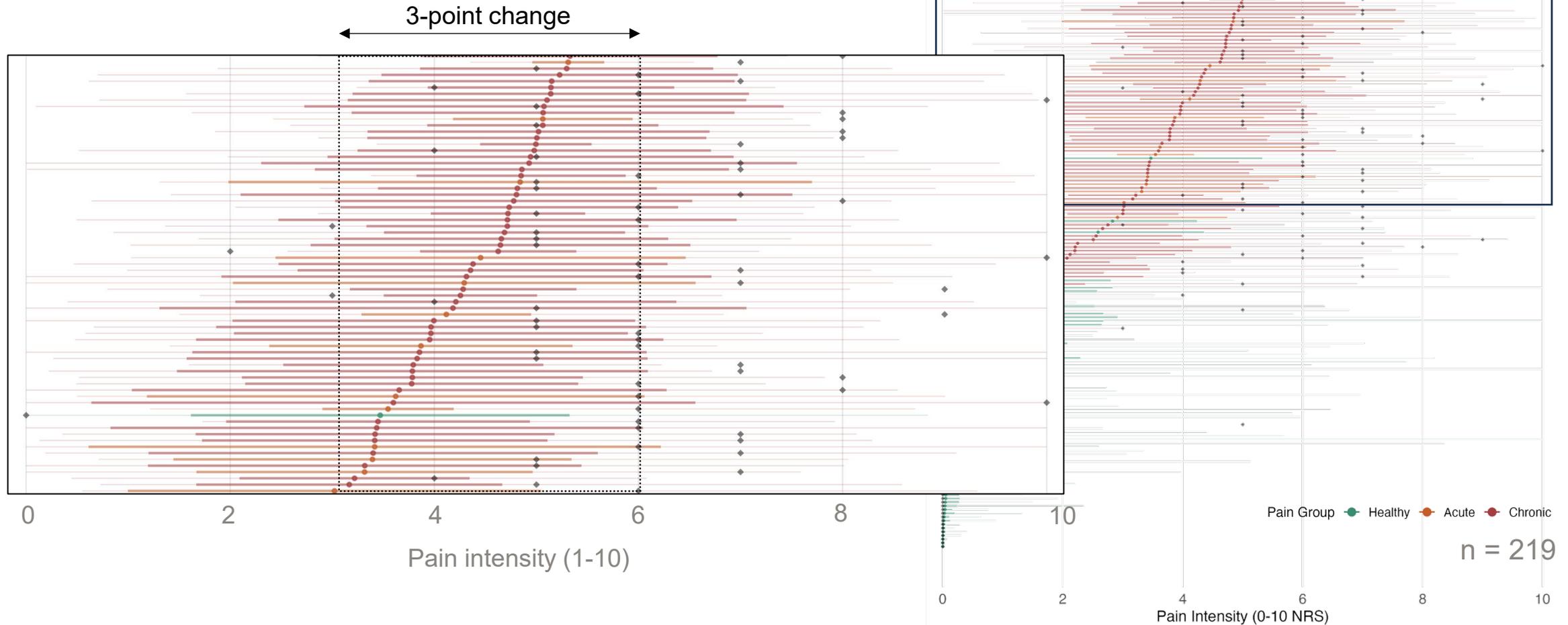


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A 3-point pain change is not a treatment signal — it's **natural fluctuation**.

Within-person pain variability alone exceeds standard responder thresholds for pain.



What we need: measures of brain **function**.

Example Computational Psychiatry

Computational measures predict better than behavior and brain imaging alone.

