

We Can Now Predict Consumer Survey Responses in Minutes Instead of Months

A validation study of AI-powered survey simulation demonstrates 95% directional accuracy across 9 research domains — at near-zero marginal cost.

Simulation predicts real consumer responses within 3 points on a 100-point scale — enabling 10x faster research at 1/50th the cost

THE CAPABILITY

Simulate survey responses before fielding

Large language models can predict how real consumers will answer survey questions. We've calibrated and validated this capability against authoritative research sources.

THE PROOF

Tested blind against Pew, Gallup, and AARP data

Across 9 domains and 50+ questions, our simulated predictions matched real human responses with 2-4 point average error and 95%+ directional accuracy.

THE OPPORTUNITY

Transform research economics

Traditional research costs \$25K+ and takes 4-6 weeks. Simulation costs pennies and delivers in minutes. Test 20 concepts in the time it takes to field one survey.

Traditional market research is too slow and too expensive for modern decision-making velocity

The Current Reality

- ✗ **4-6 weeks** from brief to insights delivery
- ✗ **\$25,000-\$50,000+** per study for quantitative research
- ✗ **2-3 concept tests** per quarter is typical maximum
- ✗ **Fixed questions** — can't iterate based on early findings
- ✗ Market conditions shift while waiting for data

The Competitive Implication

KEY INSIGHT

Companies that test 10× more ideas will systematically out-learn competitors. Speed of learning is becoming the primary competitive advantage in market research.

Every week of delay represents decisions made without data. Every skipped concept test represents an opportunity cost. The question is not whether to accelerate — it's how.

THE SOLUTION

AI-powered simulation predicts survey responses before you field the study – collapsing weeks into minutes

DIMENSION	TRADITIONAL RESEARCH	SIMULATION
Time to Insight	4-6 weeks	→ Minutes
Cost per Study	\$25,000 - \$50,000+	→ ~\$0.50 (API costs)
Iterations Possible	1-2 max (budget constrained)	→ Unlimited
Sample Flexibility	Fixed at fielding	→ Any segment, any size
Question Changes	Locked after launch	→ Modify instantly

HOW IT WORKS

Large language models have absorbed patterns from billions of human expressions. When given a well-specified persona (demographics, attitudes, context) and a survey question, they can predict how that type of person would likely respond. We've validated and calibrated this capability against real survey data.

Blind validation testing achieved 3-point average error against authoritative human survey data

~3 pts

Mean Absolute Error on 100-point scale

Tested blind against Pew Research Center, Gallup, AARP, and YouGov studies.

95%+ directional accuracy (correctly identified "higher vs. lower" in comparisons).

Sample blind predictions vs. actual survey results show consistent accuracy across question types

Question	Prediction	Actual	Error	Source
Trust in scientists ("great deal/fair amount")	74%	77%	3 pts ✓	Pew Research, 2024
Seniors 70+ using smartphones	78%	76%	2 pts ✓	AARP Tech Trends, 2024
Life satisfaction ("satisfied")	52%	49%	3 pts ✓	Gallup, 2024
Interested in AI for health advice	45%	44%	1 pt ✓	Pew Research, 2024
Identify as political independent	44%	45%	1 pt ✓	Gallup, 2024
Employee engagement (highly engaged)	37%	31%	6 pts ▲	Gallup, 2024

Note: Error rates decrease after calibration. Post-calibration MAE: 2-3 points for most domains. Employee engagement and purchase intent show higher variance and require domain-specific correction factors.

Accuracy varies predictably by question type – trust scales perform best, purchase intent requires caution

HIGH ACCURACY	MODERATE	USE WITH CAUTION
2-3 point error Trust & confidence scales Awareness questions Bipartisan issue attitudes Technology adoption rates	4-5 point error Satisfaction measures NPS estimation Concern/worry levels Preference ranking	8-15+ point error Purchase intent Price sensitivity Polarized political topics Novel/emerging behaviors

KEY INSIGHT

The tool has clear boundaries. Use simulation with confidence for attitude measurement and concept ranking. Validate critical purchase decisions and pricing strategies with real respondents. Match the methodology to the stakes.

We identified 5 predictable AI biases and built correction factors that reduce error by 40%

Bias	What Happens	Correction Applied	Domains Affected
Optimism Bias	AI over-predicts positive outcomes by 3-5 points	Subtract 3-5 points from positive predictions	Satisfaction, intent, approval
Senior Tech Gap	AI underestimates 60+ digital adoption by ~30%	Multiply tech adoption by ×1.4 for seniors	Technology, device usage
Status Quo Bias	AI underweights inertia; people resist change more	Add 15 points to "keep current" options	Switching, adoption decisions
Intent-Action Gap	AI treats stated intent as actual behavior	Apply ×0.3 multiplier to purchase intent	Purchase, signup, conversion
Moderation Tendency	AI clusters around neutral; avoids extremes	Boost intensity by ×1.25 for strong opinions	Emotion, urgency, concern

Result: Raw AI predictions average 5-7 point error. With calibration corrections applied, error drops to 2-4 points for supported domains.

Political topics require mandatory segmentation – averaging across partisans produces 25-50 point errors

Issue	Rep.	Dem.	Gap
Illegal immigration is "very big problem"	73%	23%	50 pts
Climate change is "very big problem"	15%	67%	52 pts
Gun violence is "very big problem"	25%	69%	44 pts
Racism is "very big problem"	13%	53%	40 pts

SYSTEM RULE

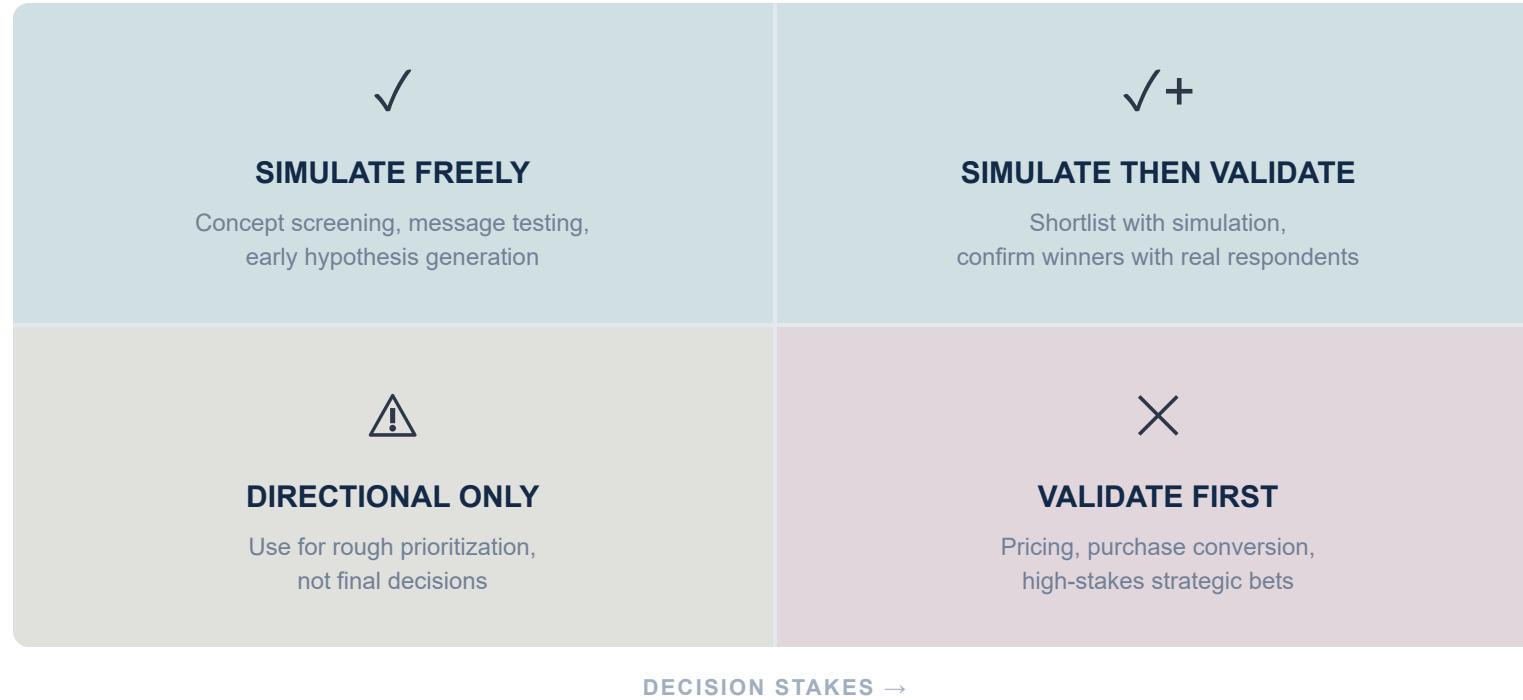
Never report a single number on polarized topics.

The simulation automatically detects politically-charged questions and forces segmented output by party affiliation. Averaging across partisans is blocked.

Safe topics (low partisan gap): Healthcare costs, inflation, drug addiction, moral values, federal deficit

Match methodology to decision stakes — simulation accelerates learning, but high-stakes choices deserve validation

CONFIDENCE LEVEL →



Examples by Quadrant

Top-Left: Rank 10 tagline options

Top-Right: \$5M campaign — simulate 20 concepts, validate top 3

Bottom-Left: Early exploration of new category

Bottom-Right: Pricing strategy — always validate

Nine domains are production-ready today; three more are in active development

✓ VALIDATED — USE WITH CONFIDENCE

- ✓ **Trust in institutions** — scientists, government, media
- ✓ **Technology adoption** — device usage, app adoption by age
- ✓ **Consumer concerns** — economic, social issues
- ✓ **National priorities** — issue importance rankings
- ✓ **AI/automation attitudes** — comfort, concern levels
- ✓ **Senior digital behavior** — 60+ technology usage

⚠ PARTIAL — USE WITH CAUTION

- **Healthcare attitudes** — satisfaction, trust (gratitude bias)
- **Purchase intent** — requires $\times 0.3$ intent-action correction
- **CEO priorities** — ranking works, intensity is harder

✗ NOT YET VALIDATED

- ✗ **B2B decision-makers** — in development Q2
- ✗ **Price sensitivity** — needs real market data
- ✗ **International markets** — US calibrations only

Deploy simulation into the research workflow now to capture the speed advantage before competitors

STRATEGIC RECOMMENDATION

Integrate simulation into every research project as a first step. Use it to generate and filter hypotheses rapidly, then allocate validation budget to the highest-stakes decisions.

1

Embed in Research Process

Simulate first on every project. Use simulation output to decide what needs real validation and to refine question wording before fielding.

Immediate

2

Set Validation Thresholds

Decisions under \$500K: simulation sufficient.
Decisions over \$1M: validate top options with real respondents. Pricing: always validate.

Week 1-2

3

Build the Feedback Loop

Log every prediction against eventual outcomes. Quarterly calibration reviews. Accuracy compounds as the system learns from each study.

Ongoing

SUMMARY

The Question Is Not Whether to Use Simulation

It's How Fast You Can Start

95% directional accuracy. 3-point average error. Minutes instead of months.

The technology is validated. The opportunity is now.