

# Customer Engagement with AI-Powered Chatbots in Retail Banking



## Background and Research Question

### Dominant Models

- Technology acceptance models (TAM, UTAUT) dominate AI-enabled service research.
- Focus on adoption and intention to use.
- Emphasise functional efficiency (usefulness, ease of use).

### Limitations in High-Risk Contexts

- AI systems are autonomous, adaptive, and opaque.
- In retail banking, customers evaluate AI beyond efficiency.
- Relational perceptions (especially trust and transparency).

### Key Research Gap

- Relational factors influence adoption, but little is known about their role in post-adoption engagement.
- Constructs studied in isolation, not jointly.

### Research Needed

- Shift focus from adoption → sustained engagement.
- Examine how multiple perceptions jointly shape engagement with AI-powered chatbots.

## Research Question

*Which customer perceptions of AI chatbots shape engagement in retail banking?*

## Methods

### Survey

- n = 30 users; measures: engagement, trust, satisfaction, empathy, personalisation; multiple regression.

### Sentiment Analysis

- ~19,000 U.S. banking reviews; logistic regression + XGBoost & BERT comparators.

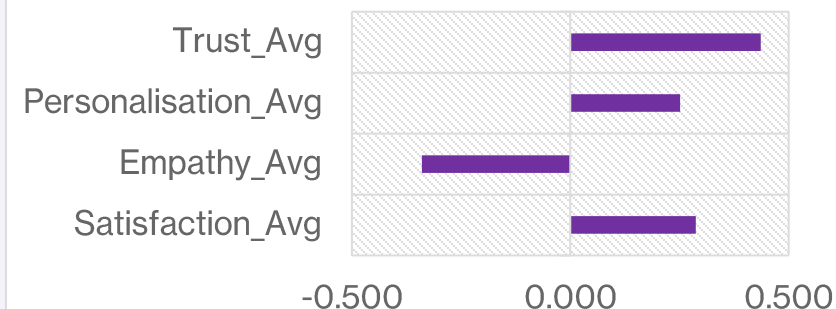
### Design

- Triangulated quantitative approach combining perceptual and behavioural evidence.

## Key Findings

- Trust strongest predictor of engagement.
- Satisfaction & personalisation weaker positive effects.
- Empathy negatively associated.
- Negative sentiment driven by functional failures which can erode trust over time.

## Relative Strength of Engagement Drivers



## Implications

**Theory:** Extends chatbot research from adoption to engagement; positions trust as central.

**Practice:** Prioritise functional reliability and trust, not artificial empathy.