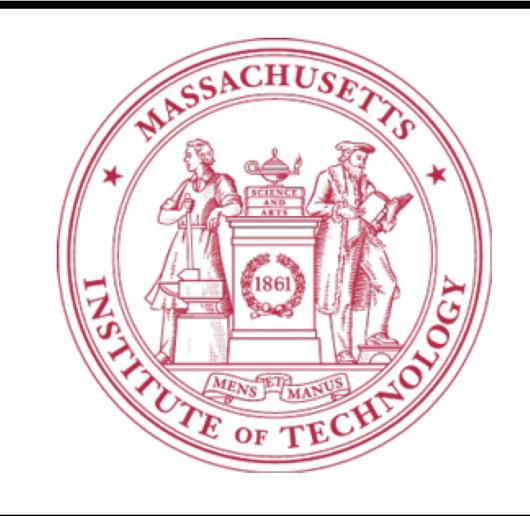


# Cognitive style impacts preference for advice seeking from Al

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#### Abstract

What kinds of individuals seek out advice generated from an algorithmic (vs. a human) advisor?

We assess the hypothesis that cognitive style, as measured through the Cognitive Reflection Test (CRT), is associated with the suppression of System 1 biases against artificially intelligent agents.

In our research, we found that greater reflection, as measured through higher scores on the CRT, was associated with greater preference for advice from an algorithmic advisor, as opposed to a human advisor.

#### Introduction

- A large body of work has identified the *features* of advisors that cause "algorithmic aversion", such as:
  - Adjustability<sup>1</sup>, accuracy<sup>2</sup>, task domain<sup>3, 4</sup>, and task difficulty<sup>5</sup>
- However, recent research has found an overall preference for advice from AI, aka "algorithmic appreciation" 6.
- Individual differences in users has been overlooked and could potentially help us make sense of the mixed literature.
- Cognitive style is a consequential individual difference:
  - Reliably relates to biases in judgments and decision making<sup>7</sup>, risk taking<sup>7</sup>, belief in fake news<sup>8</sup>, religiosity<sup>9</sup>, amongst others.

### Research question

Is cognitive style related to preference for advice from an algorithmic advisor?

 $H_1$ : More analytical individuals prefer advice from an algorithmic advisor over advice from a human advisor

#### Methodology

#### **Participants**

US-based samples from Mechanical Turk

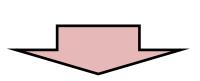
N: 183 (Study 1), 270 (Study 2), 221 (Study 3), 238 (Study 4)

#### Main measures

- 3 item Cognitive Reflection Test<sup>7</sup> (Study 1, 2, 3) or 7 item CRT<sup>9</sup> (Study 4)
- Advisor selection measure (% Advice from Al/Human)

## Design

Cognitive Reflection Test



#### Financial decision-making scenario

We want you to imagine a scenario where you have an **investment portfolio** of financial assets. You have a range of different kinds of assets and you are looking for **advice** on how to manage them all.

We will give you the option to receive advice from an **algorithmic financial advisor or from a human financial advisor**. You will be able to **choose the proportion** of advice that you receive from each of them.

# Advisor selection

Where would you like your financial advice to come from?



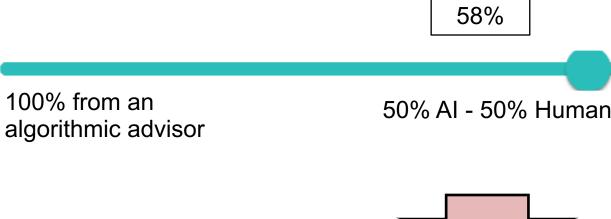
style

(7 items,

instead of 3)

Study 1: Establish relationship

between CRT & advisor choice



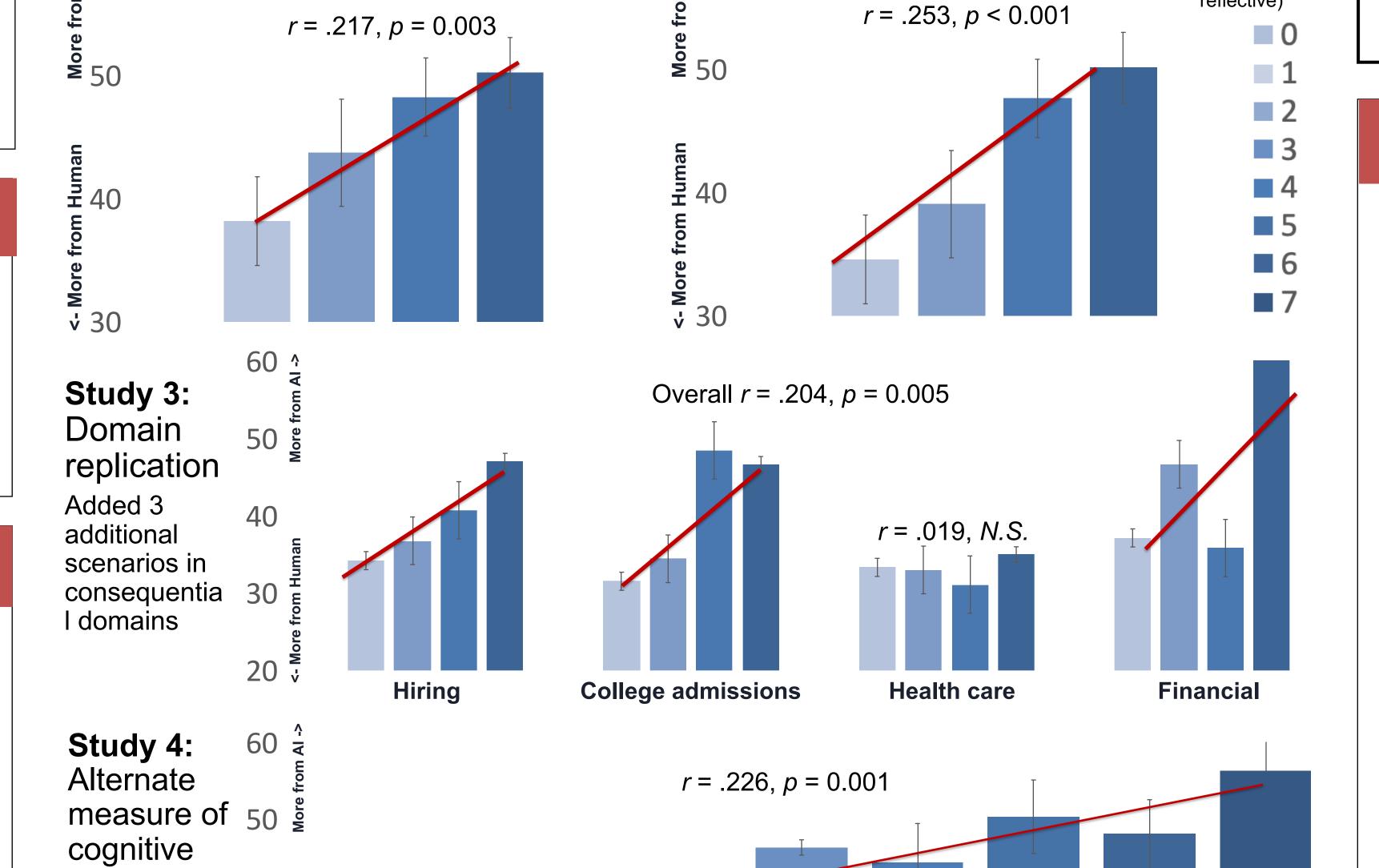


100% from a

Study 2: Straight replication CRT Score (0 = least reflective,

Demographics, controls (prior experience with AI, comfort with technology, social anxiety)

## Results



# Results summary

The more cognitively reflective individuals were, the more advice they sought from Al advisors

- Robust effect, consistently replicated (Studies 1-4)
- Across a variety of consequential decisionmaking domains (Study 3)
- Replicated using an alternate measure of cognitive reflection (Study 4)

## Implications

- Implication: Seeking advice from AI agents may be easier for those whose target audiences (consumers, employees) are naturally higher in cognitive reflection
- Alternatively, organizations wanting to increase adoption of Al advice seeking could take measures to increase reflection before decision-makers choose their advisors
- For researchers: Running studies on undergraduate samples may be biased in favor of Al technology, due to higher cognitive reflection among educated people.

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Questions or comments welcomed!
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