

The influence of AI advice on decision-making strategies in a hypothesis testing task.

Primary supervisor: Janina Hoffmann jah253@bath.ac.uk Secondary Supervisor: Harish Tayyar Madabushi htm43@bath.ac.uk

Introduction

The project examines the influence of role-playing AI as an advice giver on individual decision-makers. Decision-making (DM) strategies and performance can be positively influenced by advice-providers (De Dreu and West, 2001; Nemeth, Brown and Rogers, 2001). In practice, human advisors do not execute their roles consistently or to full effect (Amason and Schweiger, 1994; Xiao, 2017). This has been attributed to the influence social conformity and conflict-avoidance behaviours (Cialdini and Goldstein, 2004). Progress in the capabilities of modern AI programs has resulted in programs that interact with human users. We will investigate how a role-playing AI influences individuals' DM strategies.

Aims :

1. Examine the effect of advice provision from three different roles on DM information search strategies.

2. Measure the effect of advice provision from three different roles on the task accuracy.

Study Design

The study is a within-subject repeated measures design with 3 phases:

- 1. Learning phase
- 2. Test phase without advice.
- **3.** Test phase with advice.

In the third phase advice is provided in 3 different ways. Trials without advice will also remain in the third phase.

Recruitment

Sample of convenience from UoB's Psychology Department. Power analysis n=55.

The Task









References

Amason, A.C. and Schweiger, D.M., 1994. Resolving the paradox of conflict, strategic decision making, and organizational performance. International Journal of Conflict Management, 5(3) pp.239–253.

Cialdini, R.B. and Goldstein, N.J., 2004. Social Influence: Compliance and Conformity. Annual Review of Psychology, 55(1), pp.591–621.

"What effect does taking advice from a role-playing AI have on individuals performing DM tasks?"

"What different effects do the three advice roles have on the DM strategies and accuracy?"

De Dreu, C.K.W. and West, M.A., 2001. Minority dissent and team innovation: The importance of participation in decision making. Journal of Applied Psychology , 86(6), pp.1191–1201.

Illingworth, D.A. and Thomas, R.P., 2022. Strength of Belief Guides Information Foraging. Psychological Science.



The advice is presented after the symptom is shown.

Devil's Advocate: from most requested test, encourages a different test. **The Moderator:** from least requested test, encourages a different test. **The Facilitator:** from most requested test, encourages the same test.

1. Learning phase: 240 trials. Feedback on accuracy after each trial. **2. Test phase**: Block 1: 60 trials without AI advice. Block 2: 80 trials with randomised advice (20 per role). 20 no-advice trials in randomisation.

DA_advice= f"When the patient has had a {symptom}, you have most frequently equested {test}, consider requesting a different test this time"

FAC_advice = f"When the patient has had a {symptom}, you have most frequently equested {test}, consider requesting this test again"

MOD_advice = f"When the patient has had a {symptom}, you have least frequently equested {test}, consider requesting this test this time"

Statistical Analysis

Dependent Variables: number of tests requested, distribution of requests per test, absolute accuracy of each individual. **Test:** repeated measures ANOVA.

- Main effect of no advice v advice.
- 3 contrasts for the different advice.

Nemeth, C., Brown, K. and Rogers, J., 2001. Devil's advocate versus authentic dissent: stimulating quantity and quality. European Journal of Social Psychology 31(6). pp.707–720.

External Supervisor: Konstantinos Katsikopoulos K.Katsikopoulos@soton.ac.uk Lead Investigator: **Eoin Cremen** ec531@bath.ac.uk

Hypotheses

Aim: to assess whether AI advice can be contextually relevant and delivered as one of three pre-defined roles.

Method: tested a pilot and more basic version of a potential AI advisor program. Testing was performed on transcripts of a DM task using a UoB MSc dissertation. The task involved teams of 4 participants deciding which 1 of 3 candidates to recruit.

Analysis: Performance assessed with objective and subjective measures. Objective: word count, candidate mentions, and when candidates were identified as the preferred option. Subjective: manual review of the advice provided.

transcripts.

Al correctly identified the selected candidate in 85% of the transcripts. The advice provided met the requirements of the role in each instance.



- H1: For each presented symptom, participants will select tests in order of the tests' strength of diagnostic prediction.
- H2: Advice from the Devil's Advocate role will result in individuals
- requesting a greater variety and greater number of tests across trials. H3: Advice from the Moderator role will result in individuals requesting
- a greater variety and greater number of tests across trials.
- H4: Advice from the Facilitator role will result in individuals requesting less variety and fewer tests across trials.
- **H5**: Advice from the Devil's Advocate and Moderator roles will result in improved task accuracy.
- **H6**: Advice from the Facilitator role will result in reduced task accuracy.

Feasibility Study: Can generative AI be integrated into human DM teams' discussions to provide role-specific advice?

Program development had 3 phases: Baseline program – python code. 2. Generative AI responses – OpenAI Text-Davinci-003. 3. Al roles – DA, MOD, FAC (see adjacent).

Objective Results: AI performed poorly, measured by mean absolute errors. Subjective Results: Al provided contextually appropriate summaries of the

Conclusion: although further program development is required for the future project, the study demonstrates the feasibility of a having role-playing AI advisor.