Workshop at Cogsci 2021 - July 26 (Virtual)

Memory slices by Anna Strasser DISCLAIMER: JUST MEMORIES – AIMING FOR CORRESPONDENCE Meering and Revers Incering Morality WITH REALITY BUT CANNOT GUARANTEE IT.

Recent years have witnessed a burst of progress on building formal models of moral decisionmaking. In psychology, neuroscience and philosophy, the goal has been to "reverse-engineer" the principles of human morality. Meanwhile, in AI ethics, the goal has been to engineer systems that can make moral decisions, in some ways inspired by how humans do this. We aim to showcase the state of the art in both fields and to show how they can be hybridized into a computational cognitive science of morality.



- Jean-Baptiste André & Nicolas Baumard
- Shaun Nichols
- Sydney Levine, Josh Tenenbaum, Fiery Cushman
- Gillian Hadfiled
- group discussion of morning sessions

Jean-Baptiste André & Nicolas Baumard

• moral computations considering comprehensive consequences

Input of the moral system: a game defined by
(i) a vector of players' opportunity costs, C
(ii) a set of feasible outcome, Ω

Output of the moral system: a course of action for each player, given by the **Nash bargaining solution** of the bargaining problem (Ω , C)

4 IMPLICATIONS

ECTM accounts for the

- 1. logic of merit
- 2. universalization principle
- 3. apparent variability of morality
- 4. intuitions in moral dilemmas



Reverse engineering parochial morality

tribalism

- explained by in-group / outgroup & automatic group bias hypothesis
- can also be explained by rational learning
 - decision between inclusive vs. parochial norms depends on the evidence presented
- \rightarrow reduce sampling errors





with Scott Partington and Tamar Kushnir

How does moral cognition work

• FLEXIBILITY !!!



Sydney Levine

dependent rational choice.

Josh Tenenbaum, Fiery Cushmar

90-second advertisements for posters

- 1. Parker Crutchfield (Western Michigan University Homer Stryker M.D. School of Medicine) & Scott Scheall (Arizona State University), "Ignorance and Moral Judgment"
- 2. Milan Andrejević, Joshua White, Daniel Feuerriegel, Simon Laham, Stefan Bode (University of Melbourne, Australia), "Response time modelling reveals evidence for multiple, distinct sources of moral decision caution"
- 3. Enda Tan & J. Kiley Hamlin (University of British Columbia), "Probing the links between goal understanding and sociomoral evaluation in infancy using eye-tracking"
- 4. Léo Fitouchi, Jean-Baptiste André, Nicolas Baumard (Institut Jean Nicod, ENS, Paris), "Moral disciplining: the cognitive and evolutionary foundations of puritanical morality"
- 5. Cillian McHugh (University of Limerick), Marek McGann (Mary Immaculate College), Eric R. Igou (University of Limerick) & Elaine L. Kinsella (University of Limerick), "Moral Judgment as Categorization (MJAC)"
- 6. Rafal Rzepka, Yuki Katsumata & Kenji Araki (Hokkaido University), "Current Language Models Might Not Be Suitable For Reverse Engineering Moral Wisdom of Crowds"
- 7. Neele Engelmann & Michael R. Waldmann computational model of moral judge
- e R. Waldmann (Georg-August-University, Göttingen), "How to weigh lives. A of moral judgment in multiple-outcomes structures"
- 8. Sarah Wu, Tobias Gerstenberg (Stanford), "The role of counterfactual reasoning in responsibility judgments"

1. Ignorance and Moral Judgment





3. Probing the links between goal understanding and sociomoral evaluation in infancy using eye-tracking

Probing the links between goal understanding and sociomoral evaluation in infancy Ends Tan & J. Kiley Hamlin J. University of British Columbia

Background

Past behavioral research has shown that infants selectively lauch Plantin at al., 2007) and look longer of Standar et al., 2010) characters who help wirsus hinder others, suggesting that they prefer presocial (vs antisocial) characters However, the mechanisms underlying these tendencies remain under-specified.

Computational modeling research suggests that intents' perferences for protocial actors are based on the inference that prosocial actors have adopted their social pertners' unmet gools (Hamin et al., 2013) Powell 2021; Ulman, 2009

The current study ained to empirically loss the links between goal

understanding and infants' preferences for prosocial others by

- examining 5 month old intents' tooltime looking behaviors during holping/hindering events
- exploring the correlations between infants' looking behaviors particularly those related to understanding the goal of the recipient of prosocial/antisocial action/ and individual infants' holper proferences Endnesses trails

Method

The procedure and analysis plan were gooregistered on the Open Science Participants

34 intents (mean age = 5.13 months; 20 femalos) (3 holping videos + 3 hindoring videos + 1 profesence bial) = 2

Infants' looking bahaviors were necorded by an eye-tracker

Expirationsbering videos. Infants viewed the "hel" scentario power et al., 2007.



hinderer

heipe

not after viewing 6) helping and hindering Individual differences: Infants' goal-related looking behaviors predicted visual preferences for the helper infants who looked longer at the top of the bill and who showed more camber-hilling feation sequences during the climber's. ancord looked longer at the helper across preference trials the group of infants who showed these hiltop looking behaviors reliably preferred the helper to the hinderer across preference trials

Discussion

These results suggest that understanding the

goal of the dimber is important for infants" helper preferences, and support the hypothesis that infants' responses to prosocial and antisocial scenarios are based on montal state reasoning.

Main Findings

· infants showed differential eye-movement

and pupillary responses to helping and

infants showed a visual preference for the

hulper (m. hinderer) after viewing 12 (but

At the group level

hindering scenarios.



https://cic.psych.ubc.ca/research-publications/publications/

7. How to weigh lives. A computational model of moral judgment in multiple-outcomes structures

How to weigh lives. A computational model of moral judgme	nt in multiple-outcome structures.	10 10 10 10 10 10 10 10 10 10 10 10 10 1	3.0
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A Michael R. Waldmann, Department			100 100
Neele Engelmann och uter	The Long later and an and the second degree in the second se	The set of second secon	201126
 People readily make moral judgments about all kinds of outcome trade-offs in moral userment (see e.g., Awaad et al., 2018) All overarching theoretical frameworks of moral psychology acknowledge that reasoning about consequences is one important component of arriving at a moral judgment about an action (Greene, 2001, 2004; Cushman, 2013; Crockett, 2013, Mikhail, 2007, 2011) But how do we decide whether the consequences of acting outweigh the consequence of not acting in a moral scenario? 	ces		
 We propose and evaluate a computational model that predicts moral permissibility judgmen based on people's subjective utilities of consequences[action vs. consequences]inaction 		III Neele Engelmann	39
The model predicted people's moral judgments well in bour orea	man moral judgment - the part that deals with evaluating	III Neele Engeline	
 Our model could serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computational desirable to use of the serve as one building block of a complete computation desirable to use of the serve as one building block of a complete computation desirable to use of the serve as one building block of a complete com	man moral judgment of one participation of the part		

 Such a complete computational account would be desirable in its own regwhich aspects should machines imitate and which aspects should be altered

second session

"Learning from Humans to Build Moral Al" showcase a series of proposals for building ethical AI that draw insights

- how human cognition navigates the complex moral world as a starting \bullet
- place to generate engineering solutions to similar problems \bullet
 - Dylan Hadfield-Menell & Stuart Russell
 - Julia Haas
 - Alison Gopnik
 - Peter Railton
 - Henry Shevlin
 - group discussion of afternoon talks
 - Sholei Croom facilitates general discussion with questions for participants
 - additional general discussion on any topic related to the workshop

Dylan Hadfield-Menell & Stuart Russell

Takeaways



Uncertainty and cooperation are • <u>crucial</u> modeling components for learning normative properties of • the world from people



- Too many features slows down learning
- Too few leads to persistent misalignment
- Modeling (moral) learning matters

Polapec

- Imitation equilibria are fairly robust in the limit
- Pedagogic equilibria are efficient, but brittle



Artificial moral cognition

HUMANS

- moral cognition involves
 - reason (May (2018) / emotion (Prinz 2016) / hybrid (Mallon & Nichols 2011)

\rightarrow MORAL VALUATIONISM

valuation guides cognitive selection between competing moral states of affairs

e.g.: agent attributes subjective reward

- to the act or outcome of buying fair trade coffee OR
- to the determinants of the choice commodity OR
- to the idea of fair-trade practices themselves

AI

aim: a model covering a large portion of human cognition (Allen & Wallach 2012)



The biology, intelligence and computational basis of care

the social contract oriented approaches tend to ignore care as an important factor

one should consider tradeoffs

- exploitation (utility)
- exploration
- care / teaching

→ explain the foundational fact of caregiving altruism



Alison Gopnik

21:15 – Peter Railton

Ethical purport

- Ethical assessments purport to meet criteria of:
 - (a) impartiality
 - (b) generality
 - (c) objectivity and rationality
 - (d) accuracy in representing agents, actions, and outcomes
 - (e) motivation
 - (f) independence from arbitrary authority or sanction,
 - (g) intrinsic weight is given to harms and benefits to others as well as the self $_{\rm T}$
- Moral judgments can be criticized if they can be shown not to meet such criteria.



summarizing the 4 talks

Henry Shevlin,

Dylan: showed how cooperation and uncertainty help optimize outcomes. Essential if AI is going to do more than feed back our superficial desires.



Julia: A foundation to enable AI to learn contextual and fine-grained morality. Critical for AI to go beyond laws and principles that underdetermine morality. Alison: Brought out the centrality that love and relationships of care and dependency have in human affairs.

Suggests ways in which these could help create a foundation for AI morality.

Peter: A compelling account of interrelations between communication, social cognition, language and morality.

Provides the groundwork for rewarding and ethical human-AI interactions.



10 questions

(1) Alison & Peter: the 'folie à deux' problem for AI relationships.

(2) Alison & Peter: Reciprocity and dependence

Relationships are a two-way street, but does this make sense for AI?

3. Peter & Julia: Persistence of 'hot' moral disagreement.

Should we expect 'moral pillarization' in AI?

(4) Julia & Dylan: risk of decline of human moral decision-making?

Two concerns: both moral de-skilling and intrinsic value worry.

(5) Julia & Dylan: what makes valuation/reward function specifically moral?

How could an AI grasp the difference between aesthetic & moral norms?

(6) Julia & Peter: would moral AI truly be acting for moral reasons?

(7) Dylan: how far should automated 'preference' detection go?

A good friend or therapist might help us discover surprising or socially complex preferences (e.g., someone unsure about sexual or gender identity).

(8) Dylan: how to learn preferences across 'transformational' boundaries?

Should an algorithm have the option to nudge someone towards Mormonism? How about meditation?

(9) All panelists: what's the relationship between <u>being a moral decision-</u> maker and being a member of a moral community?

(10) All panelists: What would moral innovation mean for an AI? How could we distinguish this from malfunction?

thanks a lot for organizing this inspiring workshop

Organizers



Sydney Levine Harvard, Psychology MIT, Brain and Cognitive Sciences



Fiery Cushman Harvard, Psychology



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