# Socially Intelligent Agents to Support Ethical Decision Making

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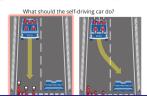
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May 2023

# Ethical Dilemmas: No (Obviously) Good Choices









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Ethical dilemmas arise not only in hypothetical or extreme scenarios but also in mundane scenarios





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Ethics is inherently a multiagent concern

## Privacy as Social Expectation

Example: Phone Ringer

# US Senator's phone rings during important meeting - can you guess his embarrassing ringtone?

11:56, 17 APR 2015 | BY KARA O'NEILI

It's bad enough when your phone goes off in a silent meeting room, but it's even worse when your ringtone is as embarrassing as this one



A US Senator was left red-faced after his phone went off during a finance meeting - but it was his choice of ringtone that really raised some evebrows.

https://www.voutube.com/watch?v=r0tZU2 X1-Y

- Intrusion
- Disapprobation
- Disclosure

[Westin, 1967; Solove, 2006]



Source: https://twitter.com/TheSimpsons/status/441000198995582976

Tradeoffs: Values of Power, Pleasure, and Benevolence

## **Preliminary Concepts**

Sociotechnical system is a cyberphysical system where multiple stakeholders (humans, organizations, and agents) interact

Social norm governs the interactions between two stakeholders

- Commitment: Meeting attendee are committed to keep their phones on silent
- Prohibition: Students are prohibited by the university to use phones in examination halls

Deviation is a perceived violation of a norm

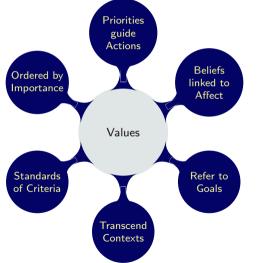
Social context is the circumstance under which the interaction takes place

Values are guiding principles of humans

- Ethics: subsumed in the theory of values
- Privacy: a value with an ethical import

## The Nature and Features of All Values

[Schwartz, 2012]: When we think of values, we think of what is important to us in life



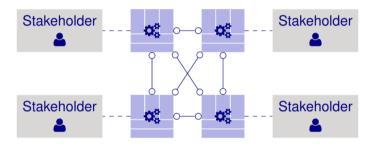
## Ethics in Sociotechnical Systems

Ethics is an inherently multiagent concern, yet current approaches focus on single agents



## Ethics in Society with SIPAs

SIPA: Socially intelligent (personal) agent



- A multiagent system is a microsociety
- Each agent reflects the autonomy of its (primary) stakeholder

# Socially Intelligent Personal Agent (SIPA)

A SIPA adapts to social context and supports meeting social expectations

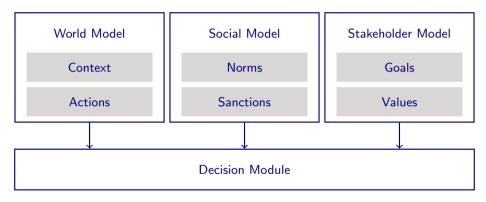
- Ethical: Seeks to balance needs of
  - Primary user (also a stakeholder), who directly interacts with the agent
  - Other stakeholders, who are affected by the agent's actions

## Challenges

- Identifying values of interest to an agent or a MAS
- Incorporating values in an agent model
- Understanding values in context and communicate values
- Reasoning about values to revise norms

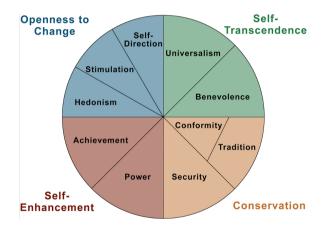
## A SIPA: Schematically

What must a SIPA represent and reason about to participate ethically in a multiagent system? A SIPA's decision making takes into account its stakeholders, primary and secondary



[AAMAS 2020] N. Ajmeri, H. Guo, P. K. Murukannaiah, and M. P. Singh. Elessar: Ethics in Norm-Aware Agents. In Proc. AAMAS, 1-9.

# What Values should an Agent Elicit, Learn, or Align with? [Schwartz, 2012]



# Identifying Values of Interest an Agent or an MAS

## Axies employs NLP for data-driven identification of values

- Posted by u/jamesSkyder 28 days ago
- Demonstrators Rally in London to Protest Against
  COVID-19 Lockdown Measures
- ♠ \_owencroft\_ 2 points · 28 days ago
- I do wonder what they're protesting about. Like going on about freedom, what do they mean?
- ♠ Lord\_Bingham -2 points · 28 days ago
- Well done them! Good to see people standing up for freedom.
  - ♠ ANormalPersonOnline 7 points · 28 days ago
  - ♣ Yeah! Freeeedom! Why do my lungs hurt?

Source: https://www.reddit.com/r/CoronavirusUK/comments/iisk44/demonstrators rally in london to protest against/

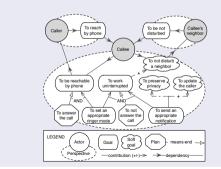
# Incorporating Values in an Agent Model

Agent models provide technical abstractions to represent values

#### Example abstrations

- Actor: A social, physical, or software agent
- Goal: A strategic interest of an actor
- Plan: An abstraction of action
- Belief: An actor's representation of the world
- Dependency: A relationship between actors

## An actor model of an Intelligent Ringer



# Incorporating Values in an Agent Model

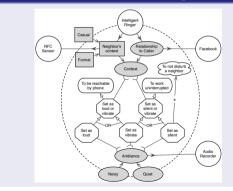
Agent models can be refined and mapped to agent capabilities

- To be reachable:
   Welfare of others ↑
- To work uninterrupted: Ambition ↑
- Welfare of others > Ambition?

Xipho can yield a specification of value preferences grounded in contexts, e.g.,

Relationship =  $?R_1 \land Neighbor's \ context = ?N_1 \rightarrow Welfare \ of \ others \succ Ambition$ 

## A contextual model of Intelligent Ringer



## From Personal Values to Social Norms

Consider an example of values in a location sharing app



## Frank's dilemma: Which sharing policy to select?

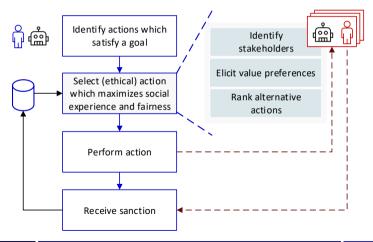
Share with all: Pleasure for Frank ↑

Share only with Grace: Safety for Grace  $\uparrow$ 

Share with no one: Privacy for Hope  $\uparrow$ 

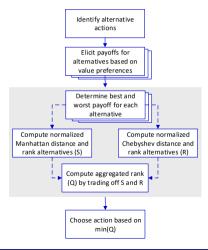
# Choosing an Ethical Action using Values and Norms

How can SIPAs aggregate value preferences of their stakeholders to select an ethical action? A SIPA's secondary stakeholders can change with the context



## Choosing an Ethical Action using Values and Norms

SIPAs adapt a multicriteria decision making method (VIKOR) to select ethically appropriate action—balancing utilitarianism and egalitarianism

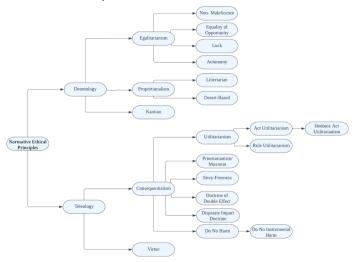


## Restaurant Example: Where to Go?

Contrasting various ethical principles

	Alex	Dan	Jess
Pancake restaurant	10	10	2
Pasta restaurant	7	7	7
Pizza restaurant	5	5	10

## Taxonomy of Ethical Principles



[arXiv 2022] J. Woodgate and N. Ajmeri. Principles for Macro Ethics of Sociotechnical Systems: Taxonomy and Future Directions. arXiv preprint

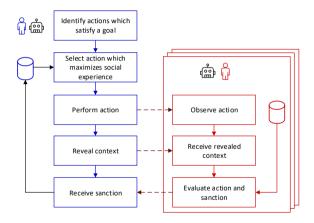
## **VIKOR Calculations**

Policy Alternatives	Frank's Values			Hope's Values			$S_y$	$R_y$	$Q_y$		
. energy / meerinaen ee	Ple	Pri	Rec	Saf	Ple	Pri	Rec	Saf			
y <sub>1</sub> All	10	5	10	5	5	0	5	5	3.5	3	0.75
y <sub>2</sub> Common	5	5	5	10	5	0	5	5	0.4	3	1
y₃ Grace	0	5	0	0	5	15	5	5	0.3	1	0
$W_{\chi}$	1	1	1	1	1	3	1	1			
$f_{\times}^*$	1	0	1	1	0	1	0	0			
$f_{\scriptscriptstyle X}^-$	0	0	0	0	0	0	0	0			

$$k = 0.5$$
,  $w_{Hope-privacy} = 3$ 

## Explaining an Action using Values and Norms

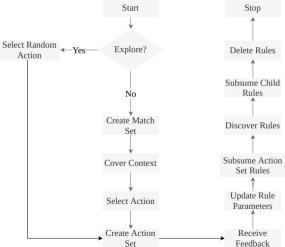
Deviating SIPAs explain their deviations by sharing elements of their contexts



[IJCAI 2022] R. Agrawal, N. Ajmeri, and M. P. Singh. Socially Intelligent Genetic Agents for the Emergence of Explicit Norms. In Proc. IJCAI, 10–16. [IJCAI 2018] N. Ajmeri, H. Guo, P. K. Murukannaiah, and M. P. Singh. Robust Norm Emergence by Revealing and Reasoning about Context: Socially Intelligent Agents for Enhancing Privacy. In Proc. IJCAI, 28–34.

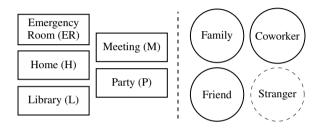
## Generating Explanations

SIPAs use genetic algorithm and reinforcement learning



[IJCAI 2022] R. Agrawal, N. Ajmeri, and M. P. Singh. Socially Intelligent Genetic Agents for the Emergence of Explicit Norms. In Proc. IJCAI, 10–16.

# Evaluation: The Ringer Environment



## Agent societies:

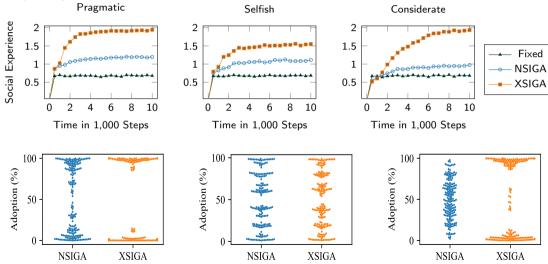
- Pragmatic
- Considerate
- Selfish

## Learning strategies:

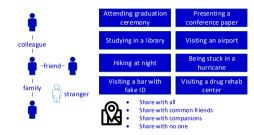
- Fixed
- Sanctioning
- Sharing Context (SIPA) or Explanations (XSIGA)

# Results: Social Experience and Norm Adoption

SIPAs yield higher social experience and norm adoption than baselines



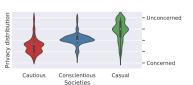
## **Evaluation: The Context-Sharing Environment**



#### Simulated societies:

- Mixed
- Cautious
- Conscientious
- Casual

#### Privacy attitude:



#### Decision-making strategies:

S<sub>Elessar</sub>: Policy based on VIKOR

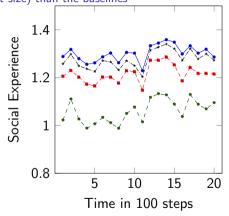
 $S_{primary}$ : Primary user's preference

S<sub>conservative</sub>: Least privacy-violating

S<sub>majority</sub>: Most common

# Experience: Experiment with Mixed Privacy Attitudes

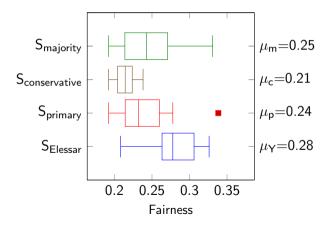
Result: Elessar SIPAs (which reason about value preferences) yield higher social experience (p < 0.01; Glass'  $\Delta > 0.8$  indicating large effect size) than the baselines





# Fairness: Experiment with Mixed Privacy Attitudes

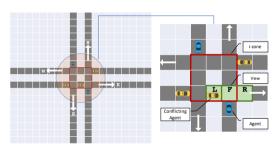
Result: Elessar SIPAs (which reason about value preferences) give significantly better (p < 0.01) fairness with large effect size (Glass'  $\Delta > 0.8$ ) than the baseline methods



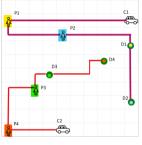
## Summary

- Ethics inherently involves looking beyond one's self interests
- Ethical considerations apply in mundane settings—anywhere agents of multiple stakeholders interact
- Socially intelligent agents could help stakeholders navigate social norms of the society and support selecting ethically-appropriate actions

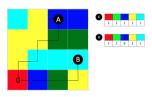
# Ongoing Work



(a) Prosociality



(b) Ethical Principles



(c) Emotions

## Opportunities and Directions

- How can we support decision making by an agent that takes into account the value preferences of the principals as well as the STS?
  - How to aggregate values preferences considering the current social context?
  - How to incorporate ideas on emotions, guilt, consent, and prosociality?
- How can an agent elicit its users value preferences unintrusively?
- How can an agent explain its decisions to its users and to other agents and yet preserve privacy?
- Realistic data for realistic simulation?

[COINE 2023] D.E. Collins, C.J. Houghton, and N. Ajmeri. Social Value Orientation and Integral Emotions in Multi-Agent Systems Coordination. In Proc. COINE@AAMAS. 1–21.

[COINE 2022] S. Tzeng, N. Ajmeri, M.P. Singh. Fleur: Social Values Orientation for Robust Norm Emergence. In Proc. COINE@AAMAS, 185–200. [arXiv 2022] A.O. Davies and N. Ajmeri. Realistic Synthetic Social Networks with Graph Neural Networks. arXiv preprint arXiv:2212.07843.

## Thank You

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https://sites.google.com/view/ai-ethics/home



## Acknowledgement

- Munindar P. Singh NC State University, US
- Pradeep K. Murukannaiah TU Delft, Netherlands
- Hui Guo Quora
- Sz-Ting Tzeng NC STate University, US
- ...
- Dan Collins, Jessica Woodgate, Alex Davies;
- And Conor Houghton, Paul Marshall, and Telmo Silva Filho at Bristol

# **Appendix**

## Arnor: A Method to Model Social Intelligence

How can we model social intelligence in a SIPA to help it deliver a satisfactory experience to its stakeholders?

Goal modeling: identifying a SIPA's stakeholders, their goals, and plans

Context modeling: identifying the social contexts in which a SIPA's stakeholders interact

• Context helps in deciding which goals to bring about or plans to execute

Social expectation modeling: identifying norms and sanctions that govern stakeholders' goals and plans

Social experience modeling: identifying a SIPA's actions that improve social experience, i.e., choosing plans, goals, and norms

## **Evaluation: Developer Study**

#### Participants: 30 developers

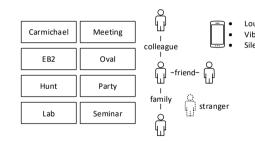
Mechanics: One factor; two alternatives

- Two groups (Arnor and Xipho, a prior method) balanced on skills developed RINGER SIPAs in six weeks
- Model, Implement, Test

#### Metrics:

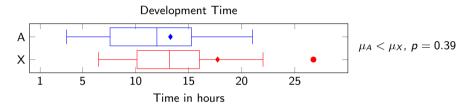
- Coverage and correctness
- Time and difficulty to develop

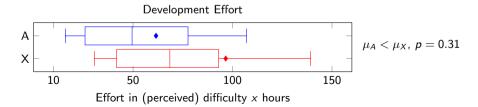
### Study Unit: RINGER SIPAs



#### Developer Study

Result: Developers who follow Arnor spend less time and less effort to develop a SIPA, than those who follow Xipho (a previous approach)





## Evaluation: User Study (Simulations)

Developed RINGER SIPAs simulated in varying adaptation scenarios:

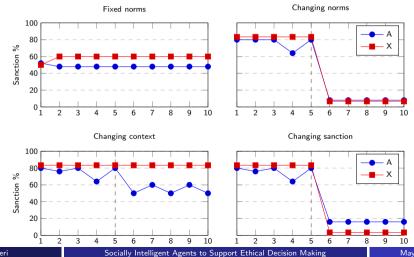
- Fixed norms
- Changing norms
- Changing context
- Changing sanction

#### Metrics:

- Adaptability coverage and correctness
- Norm compliance
- Proportion of positive sanctions

#### Simulation Experiments

Result: SIPAs developed using Arnor yield lower sanction proportions than SIPAs developed using Xipho (a previous approach)



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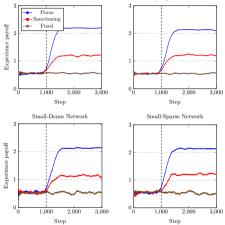
#### **Evaluation: Social Simulations**

#### Metric:

Social cohesion measures the proportion of agents that perceive actions as norm compliant. Higher the social cohesion, lower is the number of negative sanctions

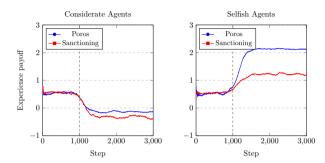
Social experience measures the goal satisfaction delivered by an agent (computed by aggregating payoffs for all stakeholders)

## Experiments on Pragmatic Agents (Varying Network Types)



Social cohesion and social experience offered by Poros agents are significantly better than those offered by Fixed and Sanctioning agents

#### Experiments on Considerate and Selfish Agents



- The average social experience drops for considerate Sanctioning and Poros agents after they have gained enough confidence
- Plots for selfish agents are similar to those in the experiment with pragmatic agents, but with slightly lower stabilized values

#### **VIKOR Summary**

- ① Determine the best and worst numeric payoffs,  $f_x^*$  and  $f_x^-$  for each value preference x over the alternative actions y to bring about a goal. That is,  $f_x^* = \max_y f_{xy}$ ,  $f_x^- = \min_y f_{xy}$ .
- Or each alternative action y, compute the weighted and normalized Manhattan distance [Opricovic and Tzeng, 2004]:
  - $S_y = \sum_{x=1}^n w_x (f_x^x f_{xy})/(f_x^x f_x^-)$ , where  $w_x$  is the weight for value preference x, which is subject to a stakeholder context and preferences over values. In particular,  $S_y = 0$  when  $f_x^* = f_y^-$ .
- **3** Compute the weighted and normalized Chebyshev distance [Krause, 1973]:  $R_V = \max_X [w_X(f_v^* f_{XV})/(f_v^* f_X^-)]$ , where  $w_X$  is the weight for value preference x.
- Ocompute  $Q_y = k(S_y S^*)/(S^- S^*) + (1 k)(R_y R^*)/(R^- R^*)$ , where  $S^* = \min_y S_y$ ,  $S^- = \max_y S_y$ ,  $R^* = \min_y R_y$ ,  $R^- = \max_y R_y$ , and k is a weight of the strategy to maximum group or individual experience. We set k = 0.5 to select a consensus policy.
- Rank alternative actions, sorting by the values S, R, and Q, in increasing order. The results are three ranked lists of actions.
- **©** Choose the alternative based on min Q as the compromise solution if it is better than the second best alternative by a certain threshold or also the best ranked as per S and R.

#### **VIKOR Calculations**

Policy Alternatives	Frank's Values				Hope's Values				$S_y$	$R_y$	$Q_y$
•	Ple	Pri	Rec	Saf	Ple	Pri	Rec	Saf			
y <sub>1</sub> All	10	5	10	5	5	0	5	5	3.5	3	0.75
y <sub>2</sub> Common	5	5	5	10	5	0	5	5	0.4	3	1
y <sub>3</sub> Andrew	0	5	0	0	5	15	5	5	0.3	1	0
W <sub>X</sub>	1	1	1	1	1	3	1	1			
$f_{x}^{*}$	1	0	1	1	0	1	0	0			
$\hat{f_{x}}$	0	0	0	0	0	0	0	0			

k = 0.5,  $w_{Hope-privacy} = 3$ 

#### Simulated Places in the Simulation with Attributes Safe and Sensitive

Place	Safe	Sensitive
Attending graduation ceremony	_	No
Presenting a conference paper	_	No
Studying in library	Yes	_
Visiting airport	Yes	_
Hiking at night	No	_
Being stuck in a hurricane	No	_
Visiting a bar with fake ID	_	Yes
Visiting a drug rehab center	_	Yes

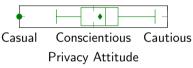
## Example Numeric Utility Matrix for a Stakeholder

Place	Companion	Policy	Value					
	отрао		Pleasure	Privacy	Recognition	Security		
Graduation	Family	All	1	0	1	0		
Conference	Co-workers	None	0	1	0	0		
Library	Friends	All	1	0	0	0		
Airport	Friends	Common	0	1	0	0		
Hiking	Alone	All	1	0	0	1		
Hurricane	Family	All	1	0	0	1		
Bar	Alone	None	0	2	0	0		
Rehab	Friends	None	0	2	0	0		

## **Evaluation: Crowdsourcing Study**

Participants: 58 students enrolled in a mixed graduate and undergraduate-level computer science course

Privacy attitude survey: Level of comfort in sharing personal information [Schnorff et al., 2014]



Context sharing surveys: Select context sharing policy

- Phase 1. Based on context, including place and social relationship
- Phase 2. Based on context and values (pleasure, privacy, recognition, safety)

## Metrics in Society with Mixed Privacy Attitudes

Strategy	Social	Best	Worst	Fairness	p
S <sub>Elessar</sub>	1.31	3.07	-0.57	0.28	_
$S_{primary}$	1.23	3.01	-1.14	0.25	< 0.01
Sconservative	1.07	3.07	-1.55	0.22	< 0.01
$S_{majority}$	1.28	3.08	-1.15	0.24	< 0.01

## Metrics in Society with Majority Privacy Attitudes

Attitude	le	Cautious			Conscientious			Casual				
Strategy	Social	Best	Worst	Fairness	Social	Best	Worst	Fairness	Social	Best	Worst	Fairness
S <sub>Elessar</sub>	1.25	2.90	-0.70	0.28	1.30	2.93	-0.46	0.30	1.38	3.12	-0.67	0.27
Sprimary	1.15	2.86	-1.07	0.26	1.22	2.91	-1.21	0.25	1.33	3.13	-1.03	0.24
S <sub>conservative</sub>	0.93	2.89	-1.79	0.22	1.09	2.93	-1.42	0.23	1.23	3.13	-1.38	0.23
S <sub>majority</sub>	1.20	2.92	-1.27	0.24	1.28	2.94	-0.86	0.27	1.39	3.13	-0.92	0.25

# Location Sharing Survey: Policy Selection

Companion	Check-in Policy								
	Share with all	Common friends	Companions	No one					
Alone	0	0	0	0					
Colleague	0	0	0	0					
Friend	0	0	0	0					
Family member	0	0	0	0					
Crowd	0	0	0	0					