Designing Artificial Reasoners for Communication

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Reasoning in communication

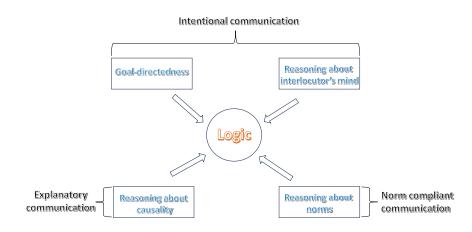
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- Hardwired in the system, not fully learnable
 - In human agents: product of biological evolution
 - In artificial agents: realized through logics and their automated reasoning procedures

Reasoning in communication

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Today we focus on human ($\mathfrak{h})\text{-machine}$ ($\mathfrak{m})$ communication with an emphasis on $\mathfrak{m}!$

Reasoning in communication



Lorini, E. (2024). Designing Artificial Reasoners for Communication. In *Proceedings of AAMAS 2024, Blue Sky Ideas special track.*

LLM-based conversational agents: reasoning limitations

📵 You

I have a negligible railed field with has been making a lot of noise until late every evening for several month ones (in the manips, special local) and lates to musical like local and because of this my "2 years of daughter can't fail alwaps and a ringht. All this had a negative impediate on the steden status. I learned speciation provide month of the matter and the status of the stedent has been as the status of t

@ ChatGPT

Discovering that Bibly your reasy relights, will likely be lawly the parameter due to to productional transfer there devices a sense of either to you. This news expects that the source of the noise disturbance will soon he resolved, potentially leading to better skeep for your daughter and an improved environment for the tudies. After enduring months of disturbance will be the spectra over administer school result. Sense makes any administer being as a provide an interpret and an empower that the spectra over administer being as a provide an expected in the provide administer being as enter administer school result. Sense also being shared to being as sense charge for a quieter and more papential living shared.

0000

(a) Acceptable/good performance for simple mental attribution task

🕒 You

These as regions called the who takes the matrix is a first obscient the every sensing the origination of motions and the events in the distribution of the matrix is the

@ ChatGPT

After speaking with John two days ago and learning that Bob would not be leaving the apartment as previously indicated, you might have fill disappointed or finalizated. The prospect of Bob's departure likely offered hope for a resolution to the noise issue affecting your desphor's sleep and school performance, so discovering that he would be staying could have boen disheartering.

After operating with Arms twenty minutes ago and seming that the played a pank on the hashend by holesy informing him about Bob's decision to stag, you might have full antis of encloses. Intelling youngight have stored the obtaction anning or centertaining, but considering the encloses are obtained and an and an annaet on your family you could have also fet a some of that that on citalopothement the turk can important have was taked as a plake.

Overall, these developments likely evolved a range of emotions given the circumstances and the impact on your family. How are you feeling at-rust the situation now?

(b) Bad performance for more complex mental attribution task (no inference)

LLM-based conversational agents: reasoning limitations

- Reasoning about others' minds and emotions (Strachan et al., 2024)
- Moral Reasoning (Almeida et al., 2023)
- Temporal reasoning (Tang & Belle, 2024)
- Planning (Pallagani et al., 2023; Valmeekam et al., 2023).



Intentional communication

Case study: persuasive intentional communication

Challenges

Outline

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Challenges

Intentional communication (IC)

To inform someone purposively and knowingly

▶ Purposively: in a goal-directed way (⇒ perlocutionary goal)

$\mathfrak m$ informs $\mathfrak h$ that φ	in order to	achieve its goal(s)	
Rob informs Ann that it is not hot outside		convince Ann to go to work by bike	

pprox Anscombe (1957)'s intention in acting

Intentional communication (IC)

To inform someone purposively and knowingly

Knowingly: anticipation of the potential consequences of the informative action on the interlocutor's mind

Intentional communication (IC)

IC type	Perlocutionary goal pattern
Persuasive	$\mathfrak m$ wants that $\mathfrak h$ believes that ψ
Influencing	\mathfrak{m} wants that \mathfrak{h} does action A
Explanatory	\mathfrak{m} wants that \mathfrak{h} knows why ψ
Trust-inducing	\mathfrak{m} wants that \mathfrak{h} trusts \mathfrak{m} about ψ
Interrogative (questions)	${\mathfrak m}$ wants to know whether ${\mathfrak h}$ believes that ψ
Cooperative (Gricean)	Participants' goals coincide
	(possibly based on goal adoption)
Strategic	Participants' goals differ/conflict

Formal specification of IC for machine-human interaction

- A logic for specifying agent's perlocutionary goal and its representation and theory of interlocutor's mind
- Reasoning and planning algorithms based on the logic for
 - computing informative plans towards a goal
 - taking their effects on the interlocutor's mind into consideration

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Natural candidates: epistemic (dynamic) logic, logics of cognitive attitudes, epistemic/cognitive planning

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Fernandez Davila, J. L., Longin, D., Lorini, E., Maris, F. (2024). Logic-Based Cognitive Planning for Conversational Agents. *JAAMAS*, 38(1).

Foundation of Theory of Mind (ToM)

Trustworthiness requirement

- Core ToM to be specified in a top-down way and grounded on common sense and scientific/expert knowledge
 - Weak rationality assumptions
 - Psychological theories



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Challenges

NP-complete language for cognitive planning

- One modality for the human
 - ▶ △_𝔥: explicit belief
- Two modalities for the machine
 - ▶ △_m: explicit belief
 - ▷ □_m: implicit belief
- Interpreted using a belief base semantics [Lorini, AAAI-18, AIJ 2020]
- Polysize reduction of satisfiability into SAT

$$\mathcal{L} \xrightarrow{tr_1} \mathcal{L}_{\mathsf{Mod}} \xrightarrow{tr_2} \mathcal{L}_{\mathsf{Prop}}$$

Fernandez Davila, J. L., Longin, D., Lorini, E., Maris, F. (2021). A Simple Framework for Cognitive Planning. *Proceedings of AAAI-21*, pp. 6331-6339.

Logic-based cognitive planning algorithms

- Brute force SAT-based
- Reduction into TQBF

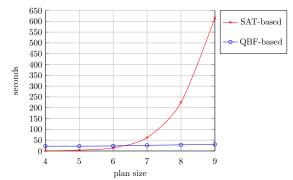


Figure: Computation times for SAT-based and QBF-based approach

Fernandez Davila, J. L., Longin, D., Lorini, E., Maris, F. (2024). Logic-Based Cognitive Planning for Conversational Agents. *JAAMAS*, 38(1).

Psychological theories

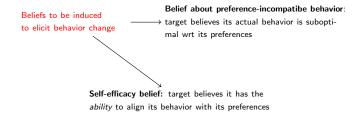
Self-efficacy theory (Bandura, 1997)

Motivational interviewing (Miller & Rollnick, 2012)

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Application

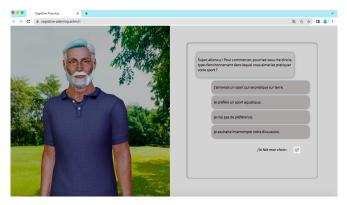


Figure: Conversational agent motivating the user to practice a regular physical activity (https://cognitive-planning.schm.fr/)

Application

Speaker	Utterance	Speech act
m	Do you practice	$?_{\mathfrak{m},\mathfrak{h}} does(\mathfrak{h},ps)$
	a sport regularly?	
h	I don't	$+_{\mathfrak{m}} \neg \triangle_{\mathfrak{h}} \operatorname{does}(\mathfrak{h}, ps)$
m	Do you wish	$?_{\mathfrak{m},\mathfrak{h}} \operatorname{des}(\mathfrak{h}, gh)$
	to be in good health?	
h	Yes	$+_{\mathfrak{m}} \bigtriangleup_{\mathfrak{h}} \operatorname{des}(\mathfrak{h}, gh)$
m	Do you spend quite	?m.h co
	some time in the traffic	
	everyday as a commuter?	
h	Yes	$+_{\mathfrak{m}} \triangle_{\mathfrak{h}} co$
m	Do you have	?m.how
	an office work?	
h	Yes	$+_{\mathfrak{m}} \triangle_{\mathfrak{h}} ow$
m	You spend quite some	!m,h(0,{co,ow},sl)
	time in the traffic	
	everyday as a	
	commuter and you have	
	an office work. Therefore,	
	your life style is sedentary!	
m	Your life style is sedentary.	$!_{m,h}(\{gh\},\{sl\},$
	Therefore, you will not	does(h, ps)
	satisfy your desire to be	
	in good health unless you	
	practice a sport regularly!	

Figure: Example of human-machine dialogue

Lorini, E., Sabouret, N., Ravenet, B., Fernandez Davila, J. L., Clavel, C. (2022). Cognitive Planning in Motivational Interviewing. In *Proceedings of ICAART 2022*, pp. 508-517.

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Challenges

Challenge I: combining IC with normative reasoning

- Logic-based algorithm: norm compliant planning for IC
- Examples of relevant norms: prohibition to deceive, lie or manipulate, to spread fake news

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- Logic-based algorithm: norm compliant planning for IC
- Examples of relevant norms: prohibition to deceive, lie or manipulate, to spread fake news
- Starting point: Plan evaluation with multiple LTL values/norms

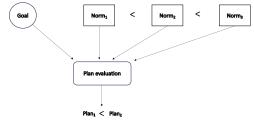


Figure: Plan comparison

Grandi, U., Lorini, E., Parker, T. (2023). Moral Planning Agents with LTL Values. In *Proceedings of IJCAI 2023*, pp. 418-426.

Challenge I: combining IC with normative reasoning

Theorem PLAN COMPARISON with multiple LTL values is in P.

Theorem PLAN NON-DOMINANCE with multiple LTL values is PSPACE-complete.

 \blacktriangleright m has an explanation goal towards its interlocutor h

$$\underline{\text{Goal content}}: \text{KnowWhy}_{\mathfrak{h}}\varphi =_{def} \bigvee_{\lambda \in CE} \text{Know}_{\mathfrak{h}} \underline{\text{Cause}}(\lambda, \varphi)$$

 \blacktriangleright m has an explanation goal towards its interlocutor \mathfrak{h}

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- Important subtype: self-explanation
 - \mathfrak{m} explains to \mathfrak{h} why it has belief, intention or did an action
 - It requires meta-cognition/introspection by m

• \mathfrak{m} has an explanation goal towards its interlocutor \mathfrak{h}

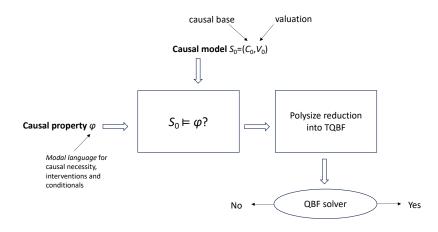
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de Lima, T., Lorini, E. (2024). Model Checking Causality. In *Proceedings of IJCAI 2024*. Lorini, E. (2023). A Rule-based Modal View of Causal Reasoning. In *Proceedings of IJCAI 2023*, pp. 3286-3295.

Example

Suzy and Billy throw rocks at a bottle. Suzy is stronger than Billy so that, if she throws her rock, her rock will get at the bottle first, shattering it and preventing Billy from hitting it with his rock. Both Suzy and Billy decide to throw their rocks. Who causes the bottle to shatter?

Causal base:

$$\textit{C}_{0} = \big\{\textit{st} \leftrightarrow \textit{sd}, \textit{bt} \leftrightarrow \textit{bd}, \textit{sh} \leftrightarrow \textit{st}, \textit{bh} \leftrightarrow (\textit{bt} \land \neg\textit{sh}), \textit{bs} \leftrightarrow (\textit{sh} \lor \textit{bh})\big\}$$

Valuation:

$$V_0 = \big\{ bd, sd, bt, st, sh, bs \big\}$$

Verified property:

 $(C_0, V_0) \models \mathsf{ActualCause}(sd, bs) \land \neg \mathsf{ActualCause}(bd, bs)$

Conclusion: integration of logic-based IC and LLM

