

Interacting with agents without a mind: the case for artificial agents

Rebecca Geiselmann^{1,2}, Afrodit Tsourgianni⁵, Ophelia Deroy^{2,3,4}, and Lasana T. Harris^{5,6}

¹ Graduate School of Systemic Neurosciences, Faculty for Biology, Ludwig Maximilian University of Munich, Großhadernerstr. 2, 82152 Planegg

² Faculty of Philosophy, Ludwig Maximilian University of Munich, Geschwister-Scholl-Platz 1, 80539 Munich

³ Munich Center for Neuroscience, Ludwig Maximilian University of Munich, Großhadernerstr. 2, 82152 Planegg

⁴ Institute of Philosophy, School of Advanced Study, University of London, Malet Street, London, WC1E

⁵ Department of Experimental Psychology, University College London, 26 Bedford Way, London WC1H

⁶ Alan Turing Institute for Data Science and Artificial Intelligence, 96 Euston Road, London, NW1 2DB

Corresponding author:

Rebecca Geiselmann, r.geiselmann@campus.lmu.de

Abstract

Humans may deprive each other of human qualities if the social context encourages it. But what about the opposite: do people attribute human traits to non-human entities without a mind, such as Artificial Intelligence (AI)? Perceived humanness is based on the assumption that the other can act (has agency) and has experiences (thoughts and feelings). This review shows that AI fails to fully elicit these two dimensions of mind perception. Embodied AI may trigger agency attribution but only humans trigger the attribution of experience. Importantly, people are more likely to attribute mind in general and agency specifically to AI that resembles the human form. Lastly, people's pre-dispositions and the social context affect people's tendency to attribute human traits to AI.

Introduction

In 2017, Sophia, a humanoid robot, was granted citizenship - a fundamental human right [1]. Five years later, one of Google's engineers reported that the company's Artificial Intelligence (AI) had become sentient [2]. Although peculiar, these events highlight that AI's abilities, rights, responsibilities, and societal roles remain ambiguous: do people consider AI as a machine or a human-like agent when interacting with it?

People predict the actions of humans based on the fundamental assumption that they are intentional agents that have a mind [3]. However, the mind is in the eye of the beholder, which means that it can be withdrawn from human agents (i.e. dehumanisation) but also ascribed to non-human agents (i.e. anthropomorphism) based on cognitive or motivational features associated with the perceiver, as well as physical and behavioural features of the perceived entity [4].

Since its beginning in 1956, AI has embraced the idea of simulating (i.e., imitating with the use of models) human intelligence, including scientific knowledge, common sense, and self-improvement [5,6]. Some see human intelligence as a property of internal thought processes and reasoning, while others focus on intelligent behaviour as an external characteristic [7]. Intended to be a machine that thinks or acts like humans, it has a growing impact on humans' social and individual lives: from almost undetectable algorithms that execute tasks on our behalf (e.g. setting prices in online markets) to those we are more aware of because they voice personalised advice (e.g. home assistant 'Alexa')[8]. Embodied AI, such as social robots, is even more noticeable and designed to interact closely with humans as helpers and companions in public places such as supermarkets, education, health care, and retirement homes [9]. Social robots do not only simulate humans in the way they think or act but also in their looks. The goal is for people to interact with them more intuitively and naturally [10].

While AI simulates a range of human-like features, its ontological status anchors them as machines or non-human agents. Nevertheless, Cockelberg (2011) pointed out that ontology matters less for society and ethics than how AI is anthropomorphised – the degree to which it *appears* to people (non-expert users) as human agent [11]. Here, we explore whether and when people consider – perceive, understand, predict, and manipulate – AI as non-human or human agent, utilising the socio-cognitive and interactive repertoire reserved for humans.

Anthropomorphising non-human entities

People predict the actions of other human agents based on the fundamental assumption that they have a mind. Daniel Dennett (1987) defined this strategy as the intentional stance and contrasted it with two more basic stances used for prediction: the physical and the design stance [12]. The behaviour of every physical system, e.g. the trajectory of a thrown ball, is subject to the laws of physics. Therefore, people can predict its behaviour by the physical stance, referring to causal-mechanical relationships. The design stance allows people to make predictions based on the assumption that systems, e.g. an alarm clock, work as they are meant to by design. In certain cases, these two strategies do not suffice: it may not be practical to predict how rational agents (e.g. humans) will behave based on these two stances. Therefore, people adopt the intentional stance, which relies on the attribution of mental states such as intentions, beliefs, or desires.

However, intentionality attribution does not necessarily imply that the other has genuine mental states in the human sense. People may also treat *non-human* entities ‘as if’ they had a mind to manage social interactions with them. This may “give us the predictive power we can get by no other method” (Dennett, 1981, p. 66) [13]. Here we argue that one of the most fundamental factors that contribute to anthropomorphism is the non-human entity’s ability to trigger the intentional stance.

The human tendency to anthropomorphise is so strong that people even readily perceive mind in animations of moving abstract shapes (e.g. disks and triangles), given they engage in self-propelled and goal-directed motion [14–16]. This tendency makes sense as being human is what people know [17]. When interacting with unfamiliar non-human entities, people may use their knowledge of themselves as a basis for understanding them. In other words, to understand the actions of these entities, people may automatically simulate similar actions in their cognitive system. Additionally, people are used to inferring other humans’ mental states to understand and predict their actions. Consequently, people may apply the same strategy to simplify unexplainable actions from non-human entities. Lastly, anthropomorphism may give people an increased sense of belonging and control in ambiguous contexts [18].

Anthropomorphism and specifically mind attribution are automatic processes that activate socio-cognitive processes in a bottom-up way, primarily driven by human-like features and biological motion [15,19–22]. Therefore, especially embodied AI (e.g. robots, androids, or avatars) may have the potential to trigger anthropomorphism, with human-like embodied AI (i.e. humanoids) having the highest potential.

Do interactions with embodied AI trigger anthropomorphism?

Self-report studies reveal that people indeed attribute mind to embodied AI [23–25] and thus anthropomorphise them. When provided with verbal descriptions of robot and human behaviours across different contexts and instructed to explain why the agent engaged in the behaviour, people used the same conceptual toolbox of behavioural explanations for human and

robot agents [26]. People even ascribed the same level of mind to humanoids and human agents when asked to rate them from different images and verbal descriptions [27].

According to Gray, Gray, and Wegner (2007) attributing mind consists of two dimensions: the capacity for agency (covering one or several of the following capacities: self-control, morality, memory, emotion recognition, planning, communication, and thought) and the capacity for experience (covering one or several of the following capacities: hunger, fear, pain, pleasure, rage, desire, personality, consciousness, pride, embarrassment, and joy) [28]. Humans are willing to treat embodied AI (here robots) as entities with some degree or kind of agency but are reluctant to perceive them as entities that can experience mental states.

This is in line with neuroscience research. Interactions with embodied AI seem to elicit action perception and representation mechanisms [29], such as motor resonance [30], motor contagion (a behavioural manifestation of action representation) [31], and the vicarious sense of agency (dependent on action representation) [32]. Action perception and representation are based on visual cues and only elicited when the other is assumed to have the capacity to act i.e. has *agency*. In contrast, research suggests that interactions with embodied AI, including humanoids, do not activate the mentalising network in the same way as humans do. Brain areas that are less or not activated by embodied AI include the TPJ [33–36], mPFC [35], and dPFC [34]. The activation of this network reflects the inference of the other's mental states and thus the assumption that the other can *experience* mental states.

Importantly, people ascribed high agency and experience to an adult. However, they attributed high experience but no agency to a baby and high agency but no experience to god. As no one would argue that babies are not human, experiencing mental states seems to be considered more ‘uniquely’ human than the ability to act [28]. Even with other humans, people tend to demonstrate lower concern for others' mental states as a function of subjective perception of distance [37]. Therefore, people likely perceive embodied AI as an agent whose actions are relevant to be predicted but yet consider it as distant, outgroup member that belongs to the group of ‘non-humans’[38]. Does the same hold true for disembodied AI (e.g. chatbots or algorithms) and what effect does the human-like form have on anthropomorphism?

Is the human model responsible for AI's anthropomorphism?

Could it be that the human-like form is fully responsible for anthropomorphism and agency attribution - meaning that, short of a human-like face, body, or motion, other AI is perceived like machines? Brain areas responsible for motor resonance (here the mirror-neuron system) also responded to non-humanoid (industrial) robots' movements [39], indicating that action representation mechanisms are also employed when interacting with robots that move in a less human-like manner. However, the degree of action representation depends on motion kinematics [31].

The degree of the AI's human-like appearance also affects people's mind attribution toward a robot. For instance, people rated a robot with a face on its screen as more “minded” compared to robots with no or a silver display [40]. People's politeness norms were also triggered more often by a humanoid robot than a non-humanoid (mechanic) robot [41], suggesting that people ascribe

more intentionality to human-like robots. This is also reflected in the action representation system: While action representation is also elicited by the movements of industrial (non-humanoid) robots' movements, the degree of action representation depends on the robot's physical appearance [42].

Coordination in more natural scenarios (here playing a musical duet together) with artificial agents (a humanoid robot compared to a computer algorithm that either committed human-like or machine-like errors), as well as social inclusion (in a ball tossing game) - a reflection of social connection and prosociality - was also influenced by how human-like the agent appears both in terms of morphological traits and in terms of behaviour [43]. Overall, this suggests that people are more likely to attribute mind, specifically agency, to humanoid than non-humanoid AI.

What about interactions with disembodied AI that does not present a visual form? Embodied AI engages visual perception, such as processing faces [44], gestures [45], and eye-contact [46] in a similar way as humans do, which facilitates action representation and prediction [29–32]. Short of a visual form, disembodied AI does not activate visual perception and action representation, possibly making agency attribution less likely. Interestingly, similar to interactions with embodied AI, interactions with disembodied AI do not activate the mentalising network across various paradigms - only humans do [47–56]. Moreover, two reviews reveal that while interactions with embodied AI do not activate mental state inference processes, they drive more engagement of some social brain regions relative to human-human interactions, such as the vmPFC [57,58]. Interactions with disembodied AI, in turn, do not lead to increased activation of any social brain regions relative to humans.

Overall, AIs are not homogenous, the distinction between humanoid, embodied, and disembodied AI reveals that people are most likely to attribute mind in general and agency more specifically to AI that mimics human appearance and motion.

Do factors beyond the human model facilitate AI's anthropomorphism?

Besides factors related to the agent and how much it activates a human stereotype [59], anthropomorphism depends on people's individual pre-dispositions [17], i.e. different personality phenotypes have different tendencies to anthropomorphise. Concerning human-AI interactions, the higher the openness to experience and the higher the level of agreeableness, the stronger the mind attribution toward the AI (here a robot). This effect was mediated by the attitudes toward AI [60].

Importantly, also social contexts and structures affect anthropomorphism, including mind attribution toward AI. Perceived in-group membership with a robot resulted in a greater extent of implicit anthropomorphic inferences and willingness to interact with robots in general [61]. Moreover, observing someone interacting socially with a robot can enhance the adoption of an intentional stance. For instance, people who 'collaborated' with a humanoid were more likely to ascribe intentions toward it after the interaction than people who did not collaborate with it [62]. The attribution of intentional traits toward a robot was also higher after social compared to non-social priming. In the social priming condition, before evaluating different types of robots, participants were told that the robots represent types of agents that they will interact with in the coming decades [60].

Culture might also shape people's perceptions of robots. Japanese participants found a robot's intervention in a moral dilemma more morally permissible than a human's intervention, whereas there was no difference among U.S. participants [63]. Lastly, social contextualisation and embodiment may also interact: People less readily saw disembodied AI as embedded in social structures than humans and explained and justified the AI's actions when solving a moral (lethal strike) dilemma differently [64].

Altogether, an interplay between the AI's appearance and motion, people's pre-dispositions, and the social context determines the degree of anthropomorphism in human-AI interactions.

Does anthropomorphism affect human-AI interactions?

Generally, treating others as agents with a mind facilitates social connection and prosocial behaviour, such as decreased cheating and increased generosity [65–68]. This also seems to hold true for human-robot interactions. For instance, attributing a mind to a robot decreased aggressive behaviour towards it [69]. Moreover, mind attribution increases the social relevance ascribed to others' actions. For example, participants follow the eye movements of a robot more strongly when they are believed to reflect intentional compared to pre-programmed or random behaviour [70–72]. However, automatic anthropomorphism or mind attribution can also have negative consequences. When an agent is difficult to categorise as human or non-human, people's response times increase and accuracy rates decrease [73,74], suggesting cognitive conflict processing. Resolving this cognitive conflict can have negative effects on performance when interacting with these difficult to categorise agents.

AI does not represent both mind perception dimensions as humans do. This is also reflected in human-AI compared to human-human interactions. Agents that display a high degree of agency, but only a low degree of experience are labeled as 'moral agents' with full moral responsibilities and the ability to show intentional behavior, in particular when it is harmful [28]. As a consequence, they are more likely to be harmed by others [75,76], denied moral rights, and judged more harshly for behaviors that lead to negative consequences [28].

For instance, people provide different moral judgments of robots and humans in the case of moral dilemmas. Robots were blamed more than human agents when they failed to intervene in a situation in which they could save multiple lives but had to sacrifice one person's life [77]. These findings seem to be independent of the participant's cultural background, as similar responses were observed in the US and Japan [63]. Harmful behaviour is also expressed towards disembodied AI: In line with prior research [78–81], across five different economic games, people were less inclined to cooperate with AI agents than with anonymous humans when it was individually but not mutually advantageous to defect. Algorithm exploitation proved to be the main driver: the effect was not driven by a competitive wish to end up better off than the AI but came from accepting the decision to act selfishly and leave the AI agent less well-off [82].

In sum, anthropomorphism can improve human-AI interactions, unless the AI is difficult to categorise as either human or machine, which takes up cognitive resources. Moreover, as people tend to ascribe agency to AI, AI is treated as a moral agent with full moral responsibilities.

Conclusion

The literature review highlights differences in mind attribution towards AI compared to humans. Firstly, it seems that AI does not fully represent both mind perception dimensions. At the current state, only interactions with human agents (adults) are attributed both agency and experience. This is also reflected in human-AI interactions with AI being treated as moral agents. Secondly, AIs are not homogenous, the distinction between humanoid, embodied, and disembodied AI reveals that people are more likely to attribute mind in general and agency more specifically to AI that resembles the human form. This also provides insights into what may increase the attribution of experience, namely an embodied form. Thirdly, beyond the human form, people's pre-dispositions and the social context affect AI's anthropomorphism. This being said it remains to be investigated whether considering AI as human will involve a shifting of the human stereotype or an error due to a cognitive limitation of how the brain enables interaction with AI.

If allowed/required:

BOX:

People understand others as entities with intentionality when attributing to them beliefs, desires, or intentions to make sense of their behaviour (i.e., intentional stance). They also perceive them as entities with phenomenal experiences attributing to them emotions, moods, or pain, which is also referred to as the phenomenal stance. Some authors [83] posit that moral concern for others emerges when we consider others as a subject of phenomenal experience. Adoption of a phenomenal stance would further promote affiliation, social interaction, and cooperation. Do people take a phenomenal stance when interacting with embodied AI?

Behavioural research suggests that people empathise with embodied AI. For example, watching a robot express fear of losing its memory and then observing it lose its memory induced more self-reported empathy than a control condition in which memory was not lost [84]. Neuroscience supports these findings [85,86]. When participants observed painful actions towards robots, the same patterns of neural activation in the ACC were found [87] that were observed when viewing pictures of painful vs. non-painful human stimuli [88–90]. Hence, people may take the phenomenal stance beyond the intentional stance when interacting with embodied AI.

Acknowledgments

We thank Jessica Brough, Jurgis Karpus, and Louis Longin for their input.

Funding: This work was supported by the European Innovation Council project ‘EMERGE’ 101070918, the Volkswagen Foundation project ‘Co-Sense’ and Bayrisches Forschungsinstitut für Digitale Transformation (bidt) project ‘Co-Learn’ (to OD) and the Alan Turing Institute, the UK’s national institute for data science and artificial intelligence (to LH).

Conflict of interest or competing interest: none.

Declaration of interest: none.

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