Assessment of autonomic response in 6–12-month-old babies during the interaction with robot and avatar by means of thermal infrared imaging.

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Abstract

From birth, infants are immersed in a social environment, often surrounded by artificial intelligent agents (AIAs). However, there is a significant paucity of work on infants' psychophysiological responses, and their related interest, when interacting with AIAs. Here, the psychophysiological responses of infants during interactions with an embodied robot and a virtual human (avatar) presented on a screen are investigated. The experimental paradigm consisted of a robot producing socially communicative gestures to babies (e.g., classic head and eye gaze movements to signal "yes," etc.) as compared to a virtual human/avatar producing socially communicative gestures (e.g., waving to signal bye-bye) and linguistic interactions such as linguistic nursery rhymes (NR) in American Sign Language (ASL). Psychophysiological responses were monitored using thermal infrared (IR) imaging technology. Indeed, such technology allows the investigation of human autonomic functions and to infer psychological and affective states in a contactless manner, thus allowing an ecological testing environment. Understanding infants' psychophysiological/emotional responses to AIAs provides important new knowledge regarding how AIAs impact infants' during the first year of life, which is a period of critical importance for human learning, especially emotional, social, and language learning and higher cognitive growth.

Keywords: Thermal imaging, infant artificial agent interaction, affective computing

1. Introduction

Virtual human and robotic agents tend to occupy a growing prevalence in daily life, with applications ranging from entertainment, training, learning, to tutorial systems [1–3]. Anthropomorphic virtual agents can be a powerful technological tool to impact motivational outcomes and facilitate learning, particularly in young children [4–7]. Especially, robots and avatars might play a primary role in this regard [8,9]. However, in order to investigate the quality of the interaction between infants and artificial intelligent agents (AIAs), it is fundamental to monitor the user's psychophysiological state [10]. To this end, contactless technologies that are able to detect psychophysiological states and preserve the ecological features of the infants-AIAs interaction are preferable. By contrast, classical technologies used for monitoring physiological parameters and autonomic nervous system (ANS) activity (e.g., Heart Rate Variability, Skin Conductance Level) typically require contact sensors, which cause manipulation of the infants' body, potentially biasing the measurement of psychophysiological/emotional states.

Advances in thermal infrared (IR) imaging technologies allow monitoring human autonomic functions and inferring psychological and affective states in a contactless manner [11–13] and have proven highly suitable to investigate human-AIAs interaction [5]. Thermal Infrared imaging is a well-known and widely used imaging technique that enables the assessment of a body's thermal distribution without any contact between the body and the sensor itself [14]. Capturing the infrared radiation spontaneously emitted by the body, the thermal camera is able to convert it into a digital map of the body's superficial temperature profile. Pointed toward a human body it allows the dynamic recoding of the skin temperature distribution in a contactless and non-invasive manner [15–17]. The modulation of the cutaneous temperature is subject to various influences, including the cutaneous blood flow, vasoconstriction/vasodilation, local tissue metabolism, and sudomotor response, all phenomenon regulated by the autonomic nervous system (ANS) which is also responsible for the human body homeostasis and the physiological responses to emotional stimuli [12]. Thermal data analysis is mainly based on linear descriptive metrics (e.g., average temperature variation from a baseline) indicative of the autonomic activation. More recently, non-linear analysis methods in both frequency and time domain have been applied to IR imaging in order to investigate the autonomic functions [18,19].

So far, thermal IR imaging has been employed for measuring facial skin temperature in infants. For instance, Mizukami et al. used thermal IR imaging as an index of mental stress in 8 to15 weeks-old infants when they were separated from their mothers [20]. Nakanishi and Matsumura investigated changes in facial skin temperature in 2 to 8 months-old infants, when they were laughing, as typical behavior of pleasant and joyful emotion [21]. Aureli et al. assessed the facial

temperature variation in 3 to 4 months-old infants, in order to explore the natural human process of attachment between baby and mother, and then the effects of non-responsiveness on the part of the mother [22].

In the present study, we conducted an original investigation of the psychophysiological correlates of autonomic nervous system responses that occurred when infants were interacting with a robot as compared to a virtual human/avatar. The data were drawn from a larger innovative project to create an effective natural language learning tool for young infants, be they hearing or deaf, during the critical periods of human learning and development, called RAVE (or, Robot, AVatar, thermal Enhanced language learning tool) [5–7]). However, entirely new scientific questions and hypotheses are addressed here, and novel analyses are conducted, which render original scientific findings. As social and environmental engagement are essential elements in infant learning [Petitto et al. [23]], RAVE used a physically present (embodied) robot in front of the baby producing socially communicative gestures involving classic head and eye-gaze movements (e.g., to signal "yes," etc.) and a virtual human/avatar on a TV screen. Here, the virtual human produced socially communicative gestures as well as linguistic interactions such as linguistic nursery rhymes (NR) in American Sign Language (ASL). Thus, the key scientific question concerned whether the two artificial agents elicit different or similar psychophysiological responses in young humans? This scientific question is compelling because changes in psychophysiological responses are indicative of differences in emotional, social, and attentional engagement. Answering this question with Thermal IR Imaging offers an important scientific advance in our ability to evaluate an AIA's capacity to effect learning in young infants, as traditional means of studying the nature of engagement in infants can be inconclusive (e.g., the examination of baby eye gaze alone).

In this study, a first-time thermal infrared imaging-based technology (with novel integration of face-tracking and computational modeling) was created specifically to investigate infants' psychophysiological responses. The dynamics of the Thermal IR facial cutaneous temperature were recorded without any contact with the infant participant, thereby making it possible to assess ANS activity and to infer the infant participants' psychophysiological states with a greater ecological validity [24]. To summarize, the main scientific question was as follows: Does the baby's interaction with the two different AIAs (robot, avatar) evoke different or similar baby psychophysiological responses?

2.   Materials and Methods

2.1 Participants

Sixteen hearing infants (ages 6-12 months) participated in this study. Parents were provided with a detailed explanation of the study's procedures, questions were encouraged, informed consent was invited and obtained. The measurements were performed in agreement with the ethical standards of the Helsinki Declaration [25] and approved by the Institutional Review Board of Gallaudet University, Washington, D.C (IRB #2617).. Participation was strictly voluntary, and parents were informed that they could withdraw at any time without consequence or penalty. Demographic information, such as age and gender, were gathered from all the participants. All data were anonymized with participant number IDs to preserve confidentiality and then entered into a database.

2.2 Experimental paradigm

Infants were positioned on their parents' laps, comfortably seated on a non-swiveling chair. The experimental design consisted of an initial period of familiarization [5–7,26]. Here, babies could interact with the environment and the experimental setup. Following familiarization, the study design consisted of two (2) conditions. C1: Robot's (RO) interaction with the baby, ending with the robot turning its head towards the TV screen (which contained the avatar) and C2: Avatar's (AV) interaction with the baby. The robot's interactive behavior towards the baby consisted of performing a series of specific head and eye movements; these included, socially communicative gesture ("yes" head nod), "surprise" gesture (head nod forward/backward, eyes wide open+blink), "peek-a-boo" gestures (eyes fully opened and eyes fully closed, plus raised/lowered head nods, as if playing "peek-a-boo"), directed eye gaze (e.g. towards baby, towards TV screen), and waking-up gesture (head bob, plus eye blinks, as if waking up). The avatar's interactive behavior towards the baby consisted of performing different types of hand movements that included socially communicative gestures (e.g., waving HI, BYE), nonaction behavioral gestures (idle pose, as if thinking), and fully linguistic productions such as Nursery Rhymes (NR) in American Sign Language, with grammatically correct linguistic sentences. Further, the ASL Nursery Rhymes had linguistic rhyme structure [Petitto et al. [27]] and age-appropriate semantic content [Petitto et al [23]]. Each condition lasted 8 seconds, with 6 repetitions. To ensure that one condition was not repeated twice in a row, which would thus result in loss of baby's attention, conditions were not randomized. The type of productions that the robot and the avatar could perform are reported in Table 1. Figure 1a shows the ecological experimental environment and setup, with the robot, the avatar on the TV screen, and the baby sitting on his father's lap.

Table 1. *List of production types performed by the robot and the avatar.*

Category (Type) of Productions	Physically Present Robot	Virtual human, Avatar on TV screen
Behavioral gestures	*Yes*	*Yes*
Communicative/ social gestures	*Yes*	*Yes*
Linguistic gestures	*No*	Yes

2.3 Thermal IR imaging data processing
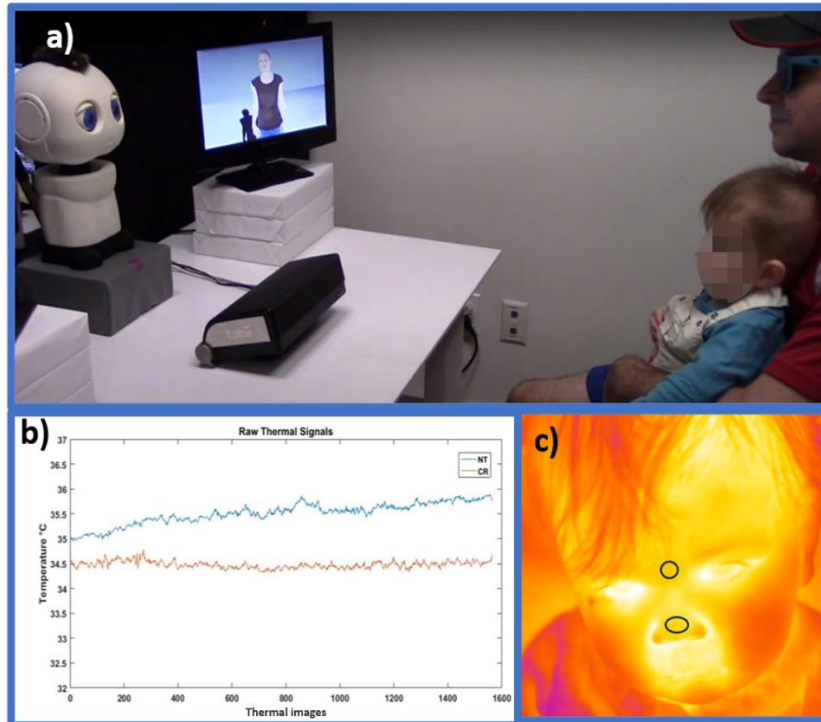
2.3.1. Data acquisition

The subject's psychophysiological response was inferred through his/her facial cutaneous temperature changes over time, recorded by thermal camera FLIR A655sc. The camera was equipped with a Focal Plane Array of 640 × 480 detectors, 0.02s time resolution, 0.03 K temperature sensitivity/Noise Equivalent Temperature Difference, and the capability of collecting the thermal radiation in the 7.5–14µm band. The thermal camera response was blackbody-calibrated in order to remove noise-effects related to the sensor drift/shift dynamics and optical artifacts [28]. The sampling rate for thermal imaging was set at 10 frames/s. All the observations were made in a climate controlled room according to the International Academy of Thermology (IACT) guidelines [29]. The room temperature was set at 23 ± 1 °C; relative humidity at 50–55% with no direct ventilation on the subject and no direct sunlight. The observations were located in a dedicated laboratory and took place at approximately the same time of the day [30]. The thermal camera was focused on the infant's face. Visible images were also recorded trough the Logitech C920 HD Pro webcam with 1920 x 1080 video resolution. The webcam was placed ensuring the co-linearity of the centres of its optic and the optic of the thermal device.

2.3.2. Data analysis

A visual inspection of the changes in facial thermal imprints in all subjects was performed to qualitatively investigate the infant's autonomic responses and to check the quality of the recording (e.g. exaggerated motion of the participant) throughout the experiment [31]. This analysis was followed by a quantitative estimation of temperature variations of Regions of Interest (ROIs), located in the infant's face (Figure 1c). In particular, to monitor the modulation of the autonomic activity, the nasal tip (NT) area and the supraorbital area associated with the activity of the corrugator muscle (corrugator region (CR)), were selected according to previous studies with infants populations [21,22]. Indeed, these two regions have been shown to be highly sensitive to the psychophysiological activity [32–35]. Because the infant was allowed to move its head without any restriction, a soft-tissue tracking algorithm was used to track the NT and CR areas throughout all the images of the time-series. The algorithm allowed to correct the ROIs positioning upon movements of the subject's head and accurately compute the ROIs temperature from each facial thermogram. The tracking algorithm, was based on a facial landmark detection over time, it was developed in Python and described in Cardone et al. [36]. In detail, it was performed based on the following processes:
-   Automatic recognition of facial landmarks in the visible domain. This task was performed using OpenFace library. In fact, on standard RGB images, OpenFace can detect facial landmarks, estimate head pose, eye-gaze, identify facial action units [37].
-   Identification of ROI with respect to detected facial landmarks.
-   Co-registration of the ROI in the visible domain with the corresponding ROI in the thermal images. A geometric transformation of the visible coordinates (calculated based on different field of view), resolution, and position was used to perform the co-registration.

In this way, it was possible to automatically extract the thermal signals in the defined regions of interests during the whole experiment. The thermal signals extracted from the 2 ROIs, were further corrected from residual motion artifacts. Motion errors were identified and replaced using a linear interpolation of the value neighboring the artifact. An example of an infant's NT and CR temperature time course after being corrected for motion artifact is shown in Figure 1b
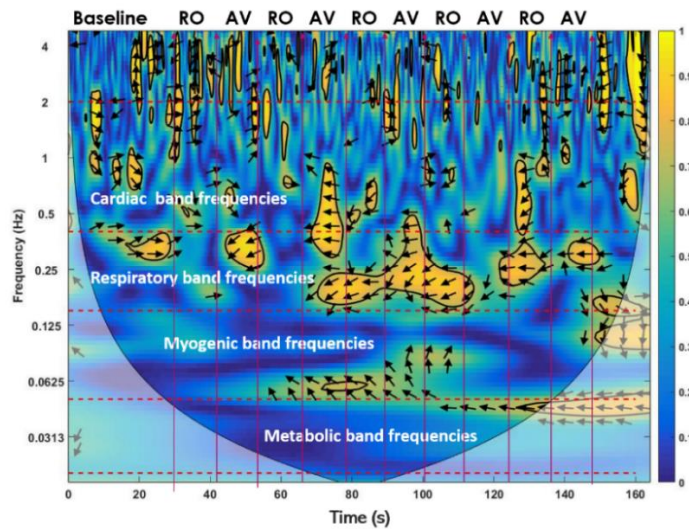
Fig.1 *(a) Experimental environment and setup with the robot the avatar on the TV screen and the baby sitting on his father's lap. (b) Temperature time course within the ROIs extracted through the tracking algorithm and corrected for motion artifacts for further analysis. (c) Example of selected ROIs (blue circle on baby's NT and CR) of a representative baby overlaid onto the baby's first frame of thermal video recording.*

Thermal data were analyzed employing both a time and a time-frequency domain approach. For the first analysis, the thermal signals were filtered with a bidirectional third-order Butterworth low-pass 2Hz filter in order to reduce high-frequency noise. The temperature time course was then normalized and averaged over each condition for each participant. The average normalized temperature during RO and AV was further evaluated with respect to the familiarization period by subtracting the average normalized temperature value of the familiarization period from each condition [28].

Concerning the second approach, in order to analyze the thermal signals at different frequencies, but without loss of timing information, a Wavelet transform was performed on both signals. In fact, thermal signals are nonstationary in nature and thermal response time in infants can be different from subject to subject. The frequency bandwidth preserved by the filter included bands in which the skin blood flow oscillations occur, corresponding to the endothelium-related metabolic (0.008– 0.02 Hz), neurogenic (0.02–0.05 Hz), myogenic (0.05–0.15 Hz), respiratory (0.15–0.4 Hz) and cardiac regulations (0.4–2 Hz) [38]. Exploring these frequency bands allowed to investigate the principal factor that contributed to the signal frequency content in the different experimental conditions. A continuous wavelet transform (Morse wavelet, gamma equal to 3 and time-bandwidth product equal to 60) was performed to investigate the frequency at which the highest signal power occurred (f_{HP}). The wavelet transform provides a multiresolution analysis by decomposing the signals into a time-frequency space. The signal power content can be thus evaluated as a function of frequency and time [39].

Finally, a coherence analysis between the 2 ROIs signal was conducted to investigate the presence of common oscillatory behavior and identify the frequency at which the highest coherence occurred (f_{HC}). The analysis was performed using a wavelet phase coherence approach [40], the coherence was computed using the analytic Morlet wavelet. This approach has been already proposed for physiological signals frequency domain analysis [28,41–43]. As with other types of transforms, the wavelet coherence applied to a finite length time-series inevitably suffers from border distortions. The region in which the transform suffers from these edge effects is called the cone of influence (COI). Similarly, to spectral analysis, errors will occur at the beginning and ends of the spectrum because of the limited time series. Therefore, the familiarization period and the sixth conditions' repetition, which were not included in the COI, were disregarded from f_{HP} and f_{HC} further analysis. Wavelet coherence values as a function of frequency and time of a representative infant is reported in Figure 2. Transparent regions outside the black line (COI) delineate regions where edge effects are significant.

Fig. 2: *Wavelet coherence between NT and CR thermal signals of a representative subject in both time and frequency domain. The black contours surround the 95% statistical significance area against red noise background assessed by Monte Carlo methods. In areas where the coherence exceeds 0.5, arrows are displayed to show the phase lag of CR with respect to NT thermal signal.*

2.4 Statistical analysis

Statistical analyses were performed using SPSS software (IBM SPSS Statistics 19). The One-way repeated measure ANOVA was computed on the conditions' (RO and AV) average temperature of both ROIs. Furthermore, it was performed on the conditions' average value of f_{HP}, and f_{HC}. The p-values were corrected for multiple comparisons using the Bonferroni adjustment. The goal of this analysis was to infer the autonomic response of the subjects and answer the theoretical question underlying the present study i.e. whether or not the baby's interaction with the two different AIAs evoke different baby psychophysiological responses.

3. Results

Regarding the time domain analysis, the one-way analysis of variance showed a significant difference in average temperature on NT between the 2 conditions ($F(1,15)=4.94$, $p=0.04$) but not for CR. The results revealed a decrease in average temperature during the avatar/AV condition with respect to the robot/RO condition.

Concerning the time-frequency analysis, both NT and CR signals showed a statistically significant difference in the f_{HP} during AV and RO (NT: $F(1,15)=7.29$, $p=0.02$; CR: $F(1,15)=9.49$, $p=0.01$). AV exhibited a stronger respiratory activity since the AV f_{HP} found on the respiratory band was statistically higher with respect to the RO f_{HP}. This result suggests a higher breathing rate during the AV compared to RO condition. This evidence was also supported by the coherence analysis results, indicating a frequency coherence peak on the respiratory band during AV condition and on the cardiac band during the RO ($F(1,15)=9.07$, $p=0.01$). The statistical results are summarized in Table 2.

Table 2. *Results of the test for differences between the 2 conditions (RO and AV). The significance has been tested trough a one-way repeated measure ANOVA.*

	RO (mean value ± standard error)	**AV (mean value ± standard error)**	**Significance**
NT Average Temperature	0,53 ± 0,02	0,47 ± 0,03	F(1,15)=4,94; p=0,04
CR Average Temperature	0,52 ±0,03	0,51 ± 0,03	F(1,15)=0,032; n.s.

NT f_{HP}	0,18 ± 0,03 Hz	0,28 ± 0,05Hz	F(1,15)=7,29; p=0,02
CR f_{HP}	0,23 ± 0,05 Hz	0,31 ± 0,05Hz	F(1,15)=9,49; p=0,01
NT-CR f_{HC}	0,53 ± 0,03Hz	0,44 ± 0,03Hz	F(1,15)=9,07; p=0,01

4. Discussion

In recent years, artificial intelligent agents have received increasing attention because of their use in smart applications that affect multiple aspects of daily life. Mostly research involving children (and to a lesser extent young infants) has focused on robots that provide social and cognitive support, the so-called "socially assistive robots." Here, children have shown their understanding of humanoid robots as being socially communicative agents [44,45]. Conversely, lesser attention has focused on virtual humans/avatars that are represented on a screen, such as a TV screen/monitor and other contemporary smart devices (e.g., iPads, smartphones) despite the fact that this usage is now ubiquitous. In fact, among the small number of studies that do exist, researchers have argued that placing young infants in front of a TV screen is judged to be an ineffectual practice [46], [47].

In the present study, we conducted a novel investigation of the psychophysiological responses of young babies during interactions with a physically present/embodied robot versus a virtual human/avatar presented on a TV screen. The overarching goal was to shed new light on the powerful contribution that advances in *Thermal IR Imaging* have as an innovative tool to provide new answers about the nature of infant social, emotional, and higher cognitive development, as well as early learning, hitherto not possible with traditional behavioral methods. Here, our use of Thermal IR Imaging—a non-invasive and contactless technology—permitted an ecological testing environment where babies were free to move and rest on a parent's lap without any major restrictions. As a further advance, we created an entirely novel infant perceptual measurement system based on thermal IR affective computing.

Using such innovations in Thermal IR Imaging, we asked whether babies would show the same or different psychophysiological responses to the robot versus the avatar. Because the robot was physically present and produced socially-communicative head and eye-gaze gestures directly to the babies—as compared to the virtual human/avatar on a TV screen, we hypothesized that the babies may show robust psychophysiological markers of attention and engagement largely to the socially potent robot (—perhaps even ignoring the flat TV screen with the avatar).

Our Thermal IR results revealed the surprising: Babies showed powerful psychophysical responses (indicative of Autonomic Nervous System regulation) *both* to the robot and to the avatar. Most interestingly, the Thermal IR evidence demonstrated that the babies' psychophysical responses were *different* for each Artificial Agent. An *increase* in nose tip temperature was observed when babies interacted with the robot, suggesting more social engagement. This finding corroborates earlier studies in child development showing that older toddlers and children can perceive robots as socially communicative [44,45], although no other team has investigated this question in infants as young as studied here (ages 6-12 months).

By contrast, a *drop* in nose tip temperature was observed when babies interacted with the avatar. This occurred in combination with a higher f_{HP} in the respiratory band not seen in the robot condition. These intriguing findings indicate that the babies were in a heightened state of attention-vigilance engagement when interacting with the avatar, as if the avatar was eliciting sustained scrutiny and tacit analysis. That a drop in nose tip temperature is associated with increased attentional engagement corroborates our earlier Thermal IR Imaging studies of adults [17, 44], and the increased respiratory response found during heightened attention as observed in adults [45]. As one novel hypothesis for future study, we suggest that the drop in nose tip temperature and increase in respiratory response—measurements made uniquely possible with Thermal IR Imaging—may serve as an innovative psychophysiological marker, not only to indicate a baby's heightened attentional state, but, correspondingly, of being cognitively poised or *"Ready to Learn."* In turn, this psychophysiological marker of a baby's state of being "Ready to Learn" may one day have broad transformative translational impact in the development of learning and teaching smart apps for use with young babies.

One intriguing question resulting from the present findings is why? Why did babies show heightened attention to the avatar? Why was this so even though the avatar was on a flat TV screen (as opposed to the physically present robot) and even though these hearing (non-sign-exposed) babies, by design, were selected for participation in the study because they would not understand the meanings of the avatar's predominant productions—specifically, the production of fully linguistic nursery rhymes in American Sign Language? While this is a question for future analysis and experiments, our previous findings with infants provide insight. Because human babies exhibit peaked attention to the maximally contrasting, rhythmic temporal patterning of spoken language during the critical period of 6-12 months, we hypothesize that these present babies' attention/fascination with the avatar was due to the avatar's production of the perceptually-salient (but meaningless) *patterns of language* being produced by the avatar, and more: The babies were attending to the language patterns, and most remarkably here, attending even though the language patterns were produced on the hands in American Sign

Language as opposed to being produced on the tongue in spoken language. Surprisingly, the findings suggest that this stage of human language acquisition occurs universally across all world languages, be they signed languages or spoken languages (see especially Petitto et al. [48]).

5. Conclusion

The present results provide compelling evidence of how the use of Thermal Infrared Imaging can be a powerful tool to investigate human-artificial agent interactions, even when studying challenging populations, such as young babies ages 6 to 12 months old. The study revealed for the first-time insights into infants' psychophysiological responses to intelligent artificial agents, and, remarkably, the subtypes of psychophysiological responses that each of these agents elicited. Of special scientific interest, we found evidence of engagement in very young infants that was more socially based (as seen in the babies' interactions with the robot) versus engagement involving more sustained attention and vigilance (as seen in the babies' interactions with the avatar).

Future studies are needed to further investigate a wider range of psychophysiological states associated with corresponding emotional, social, and attentional responses in babies when interacting with different types of AIAs. This is needed especially in light of the intriguing implications raised here regarding the potential role of artificial avatar agents to facilitate learning in very young human infants. Further study is needed to assess how the specific content of the artificial agent's communicative exchanges contributes to the different psychophysiological responses that were observed (for example, the avatar's linguistic nursery rhymes and socially communicative gestures as compared to the robot's different socially communicative gestures). To this point, future studies most certainly will include study of deaf infants being exposed to signed language. However, here, as a first step, our use of hearing non-signing infants was one unique design feature of the present study. Had the present babies understood the signed language ASL, then the differential psychophysiological Thermal IR responses that we observed between the robot and the avatar could be attributed merely to the babies' understanding of— and therefore robust vigilant attention to—the semantic *meanings of language*. Moreover, a valuable contribution to science involves the integration of Thermal IR Imaging that measures psychophysiological processes associated with emotional engagement and attention with other neuroimaging technologies, such as functional near-infrared spectroscopy (fNIRS) brain imaging that measures changes in cerebral blood flow associated with higher cognitive processing, thereby rendering a powerful and innovative tool to lay bare the relationship between emotion and cognition at the core of human learning. Presently, we are using this thermal-fNIRS brain imaging integrated technology to gain even greater insight into the cognitive, emotional, and social factors underlying infant learning and higher cognitive development, and how these factors relate to maturational stages of brain growth in very young babies.

In conclusion, the present work paves the way for a novel procedure to gain vital psychophysiological inference during infants-AIAs interactions that have the potential to render revolutionary advance across multiple applications, such as research, and, crucially, assistive application platforms and frameworks to advance human learning in infants and across the lifespan. Especially exciting, understanding infants' psychophysiological states as associated with emotional, social, and cognitive responses to artificial intelligent agents may provide important new knowledge regarding how they can facilitate learning and higher cognitive, social, and emotional growth in young infants during the first year of life, and beyond.

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