

# “I See What You Feel”: An Exploratory Study to Investigate the Understanding of Robot Emotions in Deaf Children

Carla Cirasa <sup>1</sup>, Helene Høgsdal <sup>2</sup> and Daniela Conti <sup>1,\*</sup>

<sup>1</sup> Department of Humanities, University of Catania, 95124 Catania, Italy; carla.cirasa@phd.unict.it

<sup>2</sup> Regional Centre for Child and Youth Mental Health and Child Welfare—North, UiT The Arctic University of Norway, 9019 Tromsø, Norway; helene.hogsdal@uit.no

\* Correspondence: daniela.conti@unict.it; Tel.: +39-0957102362

**Abstract:** Research in the field of human–robot interactions (HRIs) has advanced significantly in recent years. Social humanoid robots have undergone severe testing and have been implemented in a variety of settings, for example, in educational institutions, healthcare facilities, and senior care centers. Humanoid robots have also been assessed across different population groups. However, research on various children groups is still scarce, especially among deaf children. This feasibility study explores the ability of both hearing and deaf children to interact with and recognize emotions expressed by NAO, the humanoid robot, without relying on sounds or speech. Initially, the children watched three video clips portraying emotions of happiness, sadness, and anger. Depending on the experimental condition, the children observed the humanoid robot respond to the emotions in the video clips in a congruent or incongruent manner before they were asked to recall which emotion the robot exhibited. The influence of empathy on the ability to recognize emotions was also investigated. The results revealed that there was no difference in the ability to recognize emotions between the two conditions (i.e., congruent and incongruent). Indeed, NAO responding with congruent emotions to video clips did not contribute to the children recognizing the emotion in NAO. Specifically, the ability to predict emotions in the video clips and gender (females) were identified as significant predictors to identify emotions in NAO. While no significant difference was identified between hearing and deaf children, this feasibility study aims to establish a foundation for future research on this important topic.

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**Keywords:** deafness; empathy; emotion recognition; human–robot interaction; humanoid robots; NAO

## 1. Introduction

Human–robot interactions (HRIs) have been extensively explored over the last decade, contributing to a nuanced comprehension of how humans and robots engage in diverse environments (for a comprehensive review, see: [1]). Goodrich and Schultz [2] state that all interactions between humans and robots require some type of communication. Moreover, speech and language are considered to play a vital role in HRIs [3]. As robots increasingly interact with humans in various areas of their lives, understanding how people perceive emotions in robots is crucial for advancing the research in the field, as well as to improve interactions between humans and robots across different domains. Understanding robot emotions can be achieved by examining the robot’s body posture, tone of voice, or facial expressions [4]. Over the past decade, studies have placed significant emphasis on ensuring that robots not only comprehend human emotions, but also respond to them in a personalized manner [5]. This implies that robots must be advanced and capable of conducting detailed analyses of human expressions. Simultaneously, they should possess the ability to distinguish variations in expression, such as those between sexes [6].

While research into the potential of HRIs has made substantial progress, the capacity of humans to read emotions in robots remains an evolving domain. It is particularly crucial to advance research on the ability of children with special needs to interpret emotions in robots, especially among deaf children. For deaf children, effective interactions with the robot relies on non-verbal communication, as they are unable to hear sounds and voices from the robot. Therefore, it is crucial that the robot can convey its emotional states without sounds. In other words, making the robot's expressiveness as intelligible and direct as possible, without sound, can enable effective and positive engagement between deaf children and the robot. Thus, including deaf children in HRI research can offer profound advantages, that is, the inclusion of this group not only prevents these children from being sidelined in the swiftly evolving domain of robotics, but it can also provide researchers with a more expansive perspective on HRIs when auditory communication is not an option. However, to the best of our knowledge, there are no studies investigating the ability of deaf children to recognize emotions in robots.

This feasibility study aims to establish the groundwork for further research on interactions between robots and deaf children, by including both deaf and hearing children. Initially, the credibility of a humanoid robot without sound and speech are investigated. Additionally, the ability to recognize emotions in the robot without sound and speech is examined by omitting this feature in a controlled experimental context. Finally, the impact of empathy on the ability to recognize emotions in a robot without sound and speech are investigated.

Specifically, the following research questions are examined:

*(RQ1) How do hearing and deaf children assess the credibility of a humanoid robot without sound and speech?*

*(RQ2) Are children able to recognize emotions in a humanoid robot when they are exhibited through non-verbal communication?*

*(RQ3) Can empathy affect the ability to recognize emotions, expressed without sound or speech, in a humanoid robot?*

*(RQ4) Are the children who see a humanoid robot respond to a video in a congruent manner better able to recognize the emotions in the robot compared to children who see a robot respond to a video in an incongruent manner?*

The remaining parts of the paper is organized as follows: Section 2 reviews the scientific literature that contextualizes this work, focusing on social robots, deaf children, and empathy; Section 3 presents the method and procedure for the experiment. Section 4 present the results of the experiment, while in Section 5, the findings are discussed.

## **2. Related Works**

### *2.1. Social Robots*

Bartneck and Forlizzi [7] define a social robot as an “autonomous or semi-autonomous robot that interacts and communicates with humans following the behavioral rules expected by the people with whom the robot intends to interact”. The humanoid appearance of social robots facilitates interaction, as demonstrated in a systematic review by Sarrica [8]. Social robots are capable of both verbal and non-verbal communication, with the latter being particularly relevant for them to interact with deaf children. A study by Zinina et al. [9] revealed that users prefer social robots that utilize gestures, head movements, eye contact, and mouth animation, as opposed to robots with stationary body parts. Other studies have also showed that expressive robots, capable of displaying human-like emotions, are perceived as friendlier and more human-like, resulting in greater engagement and more pleasant interactions [10,11]. Further, humanoid social robots are stimulating for oral dialogue; they have been found to better facilitate language development compared to devices without human-like characteristics (i.e., tablets or mobile devices [12,13]). The potential advantages of a social robot

compared to a tablet have also been demonstrated in relation to performance, involvement, and fun [14]. Their humanoid characteristics enable robots to be able to stimulate suitable learning situations; they have been proven to be effective tutors in various settings, including second language acquisition [14]. Humanoid social robots can also enhance concentration and academic performance in both children and adolescents [15]. However, a study has shown that children's engagement and motivation in response to robotic narration are positively correlated to expressive behavior in the robot [16], indicating that robots must be able to express themselves in a specific way in order to achieve their full potential. However, more research is needed on these factors among children in general, but especially among deaf children.

## 2.2. Children with Deafness

From infancy, the sensory system acquires the ability to discern emotional signals transmitted by caregivers through various communication channels. This information can be received through, for example, facial expressions, tone and volume of voice, and body postures [17]. When one of the communication channels is weakened, other communication channels will be used to compensate for the deficiency. Hearing is pivotal in the developmental process of every child [18], when this sense is weakened, as it is in deaf children, non-verbal communication becomes fundamentally important. Deaf children and children with hearing impairment are twice as vulnerable to developing emotional and general health problems compared to children with normal hearing [19]. Further, some studies have revealed that deaf children exhibit lower levels of empathy and encounter more peer-related challenges than hearing children [20–22]. Netten et al. [23] state that because children without the ability to hear interact with others based on non-verbal communication, empathy is important for development as it can increase the ability to read other people's states and emotions in interactions.

## 2.3. Empathy

Empathy can be referred to as the ability to perceive the inner state of others as well as the ability to feel what others feel [24,25]. Specifically, empathy refers to the ability to react with the right feelings to an event [26] and is important for functioning well in social interactions and refraining from displaying antisocial behavior. Because empathy refers to the ability to feel what others feel, it is also important for emotion recognition. A study conducted by Ramachandra and Longacre [27] found that higher levels of empathy in individuals lead them to better recognize facial and eyes-only emotions. A study by Charrier et al. [28] showed that empathy influenced how empathic and intelligent a robot was perceived, as well as perceived familiarity and comfortability toward the robot. Beck et al. [29] found that body and head positions of robots can be used to convey emotions among children in human–robot interactions. However, a scoping review [30] found that most studies investigating elements related to emotions in human–robot interactions rely on mechanisms that include sounds from the robot (i.e., tones, volumes, and speed of speech), and more research is needed to better understand how emotions in robots can be interpreted by young people without sound.

## 3. Materials and Methods

### 3.1. Participants and Recruitment

Children aged between 6 and 11 were involved in the study. The sample was recruited using two different approaches. Specifically, (1) hearing children were recruited through a primary school in Catania, Italy, while (2) the deaf children were recruited from the centers of the Association of Hearing Impaired Families in Catania, Italy. A total of 9 deaf children, participated in the study. Children with comorbidities were previously not included in the sample due to the nature of the study.

All the parents signed consent forms before their children were included in the study. Children were free to leave the experiment at any time and they were always supported by a teacher or educator.

The final sample consisted of  $N = 27$  children, of which  $n = 11$  were males (40.7%) and  $n = 16$  were females (59.3%). The children had an age range from 6 to 11 years (Mage = 7.59, SD = 1.5). Within the total sample,  $N = 9$  children (33.3%) were deaf and only  $N = 2$  (7.4%) had previous experience with a humanoid robot.

### 3.2. Stimulus Material

The social humanoid robot NAO was used in the experiment. NAO is a 58 cm tall robot weighing 4.3 kg. The robot can express gestures with 25 degrees of freedom (4 arm joints; 2 for hands; 5 for each leg; 2 for the head; and one for control the hips). In addition to having several human characteristics, NAO has a light and compact design. The robot can obtain information about the environment using sensors and microphones. Further, NAO can detect faces and simulate eye contact by moving its head. The robot's eyes consist of LED lights, which can change to simulate different emotions. NAO has been shown to be suitable to use in various research settings, including experiments involving children [31,32]. According to Robaczewski et al. [33], more than 51 publications and over 1895 participants have used NAO in various types of research in 2021. This makes it among the most used humanoid robots in HRI research.

From a technical point of view, the robot was programmed with the software Choregraphe and Python 2.7+ SDK (H25, V4). We used NAOqi 2.5 framework's SDK to write Python scripts that ran on the robot and to handle its behavior during the administration.

Although NAO is not able to convey emotions using facial expressions, the program provides the possibility to design complex behaviors (i.e., different interfaces for non-verbal communication, such as gestures, sounds, and LEDs). Zinina et al. [9] found that a robot's appeal is influenced by its active features, with the most important being hands, followed by mouth, head, and eyes. NAO's emotions were programmed in line with previous studies [34,35]. Specifically, the emotion happiness was portrayed by NAO raising his hands to the ceiling and looking up (Figure 1), sadness was portrayed by NAO looking down at the floor and having his arms extended down his side (Figure 2). Anger was portrayed by NAO clenching his fists in front of him and glaring at them (Figure 3).

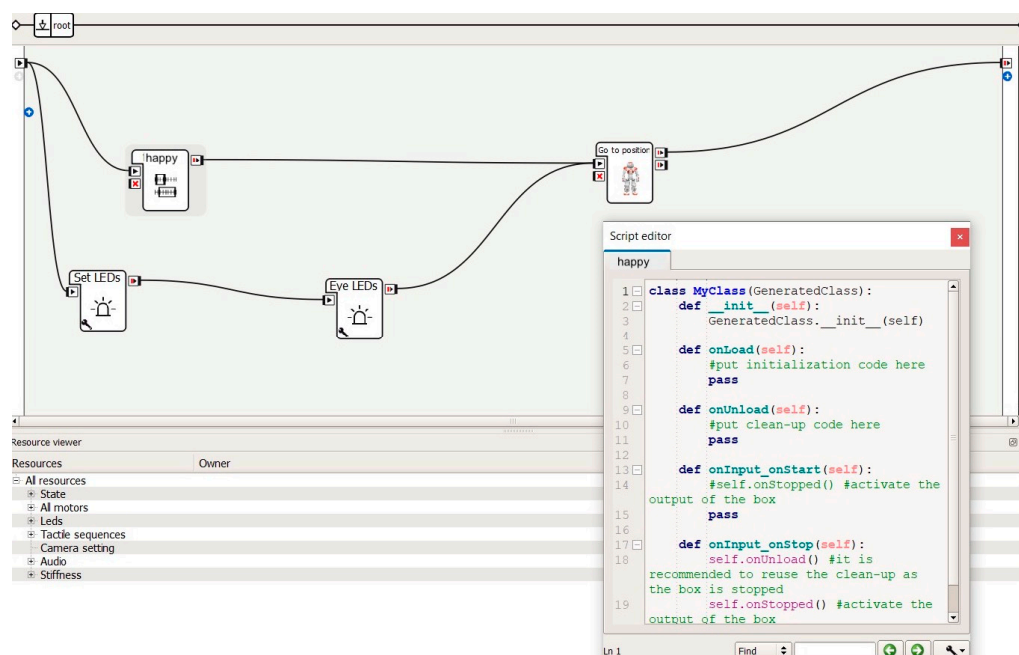


Figure 1. Choregraphe program executing the emotional reaction “happiness”.

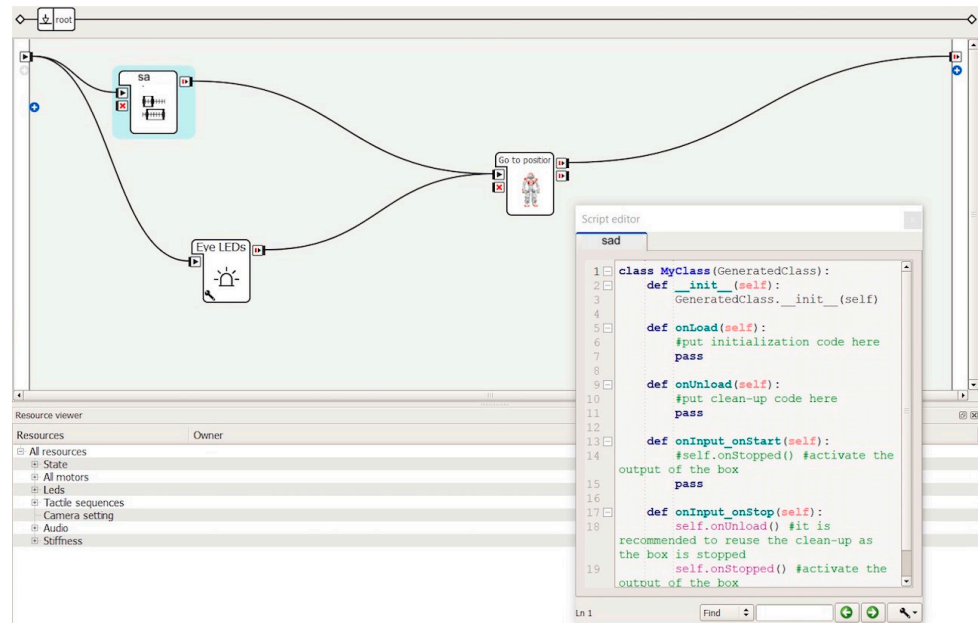


Figure 2. Choregraphe program executing the emotional reaction “sadness”.

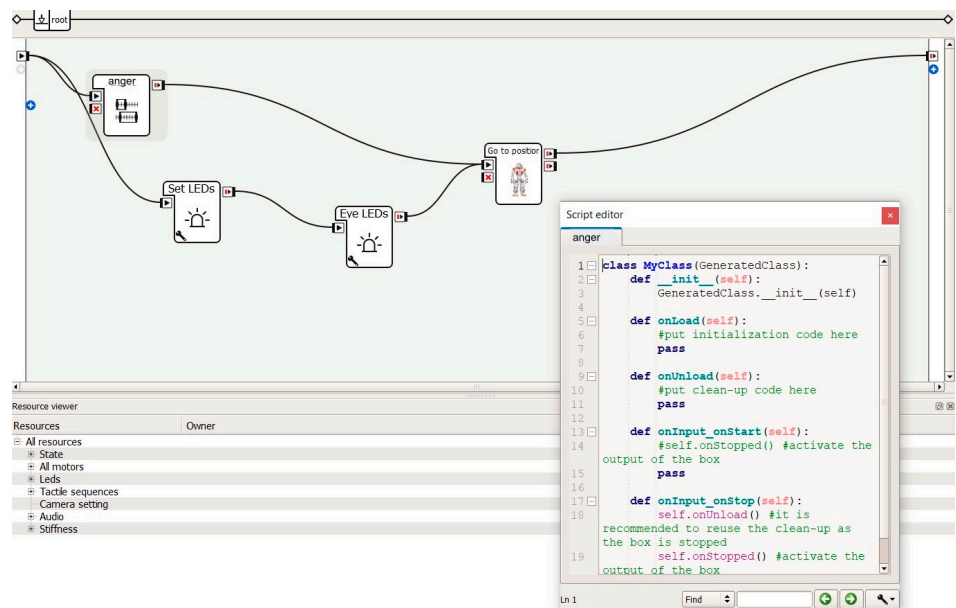


Figure 3. Choregraphe program executing the emotional reaction “anger”.

### 3.3. Instruments

#### 3.3.1. Emotions in Video Clips

Inspired by an experiment conducted by Tsiourti et al. [17], the children were presented with a list of 11 emotions (amusement, anger, disgust, despair, embarrassment, fear, happiness, neutral, sadness, shame, and surprise). The children were then asked to watch three short video clips from well-known Disney movies in random order, that depicted situations portraying the emotions happiness, sadness, or anger. Before the experiment, to favor the consistency of the instrument used, a pre-test was carried out among 5 children (2 girls and 3 boys, *Age* = 8). The videos were shown, and the children were asked “*what feeling do you associate with this video?*”. All the children independently agreed that the emotion shown in the video was in line with the emotion that was intended.

### 3.3.2. Emotion Recognition of NAO Robot

Inspired by the experiment by Tsiourti et al. [17], NAO reacted to the videos by displaying the emotions happiness, sadness, and anger. The children were asked to identify the emotion NAO displayed in line with the list of 11 emotions (amusement, anger, disgust, despair, embarrassment, fear, happiness, neutral, sadness, shame, and surprise).

### 3.3.3. Empathy

The children answered to the Empathy questionnaire for children and adolescents (EmQue-CA; [36]). The questionnaire consists of 14 items to measure empathy (e.g., “If my mother is happy, I also feel happy”; “When a friend is angry, I tend to know why”; and “If a friend is sad, I like to comfort him”). All responses were rated on a 3-point Likert scale (1 = not true, 2 = sometimes true, 3 = often true). The questionnaire has been shown to have good psychometric properties among various study samples from the age of seven [35,36], and the instrument has been translated and validated to use on Italian children [37]. Mean and Cronbach alpha are presented in Table 1.

### 3.3.4. State Empathy

To measure state empathy, we administered the State Empathy Scale test [38]. The questionnaire consists of three subscales, each with four items measuring three different dimensions of state empathy: affective empathy (e.g., “The character’s emotions are genuine”), cognitive empathy (e.g., “I can see the character’s point of view”), and associative empathy (e.g., “I can relate to what the character was going through in the message”). The scale has previously been used among different study samples, including college students [38]. Mean and Cronbach alpha are presented in Table 1.

**Table 1.** Overview of constructs and Cronbach alpha.

Constructs	No. of Items	M (SD)	Cronbach Alpha
Empathy	18	2.70 (0.54)	0.84
State Empathy (total)	12	3.62 (0.89)	0.86
State Empathy (affective)	4	3.59 (1.09)	0.79
State Empathy (cognitive)	4	3.99 (0.72)	0.43
State Empathy (associative)	4	3.27 (1.30)	0.82

### 3.3.5. Credibility toward the Humanoid Robot

The children were asked to fill out the seven questions adapted by Tsiourti [17] to assess the perceived credibility of the robot. The questions included asking their perspective on how reliable the robot reacted to emotions shown in the video clips (e.g., “The robot perceived the content of the movie clip correctly”; “The behavior expressed by the robot was appropriate for the content of the movie”) and how easy it was to recognize the emotions NAO expressed (e.g., “It was easy to understand which emotion was expressed by the robot”) as well as the anthropomorphic features of the robot (e.g., “The robot has a personality”; “It was easy to understand what the robot was thinking about”).

### 3.4. Experimental Procedure

The experiment was conducted in two different settings (i.e., a primary school or in a center for deaf children). The participants were randomized through a simple randomization (i.e., subject 1 belonged to the congruent condition, subject 2 belonged to the incongruent condition). The teacher in the class determined the order in which the students were to carry out the experiment. The conditions were blind to the children participating in the experiment. All parts of the experiment took place in a room with good lighting conditions. The rooms consisted of one computer screen, a NAO robot, and two researchers. The deaf children also had a sign language interpreter with them, who

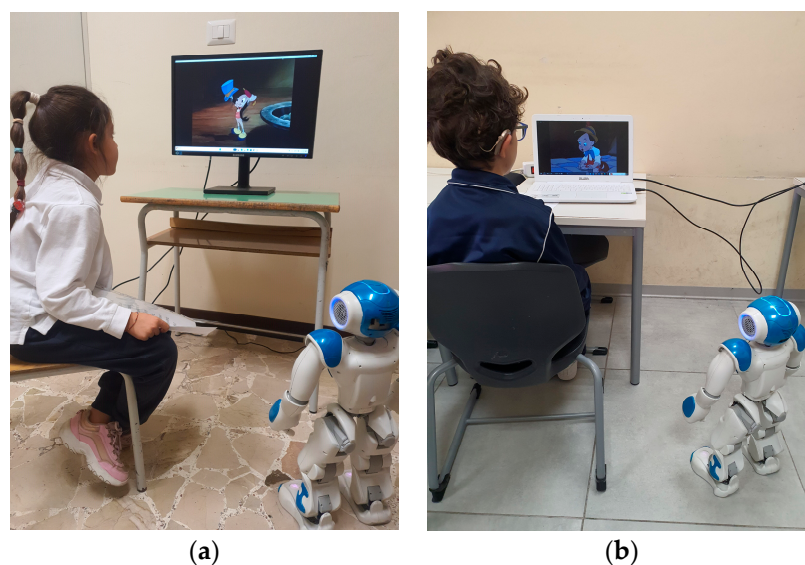
could communicate with the children on behalf of the researchers. Before the experiment, all children interacted with NAO for 20 min (see Figure 4) to minimize the possible novelty effect. The experiment was conducted without any voices or sounds, lasted for approximately 30 min, and were divided into three sessions.



**Figure 4.** Example of first meeting with NAO.

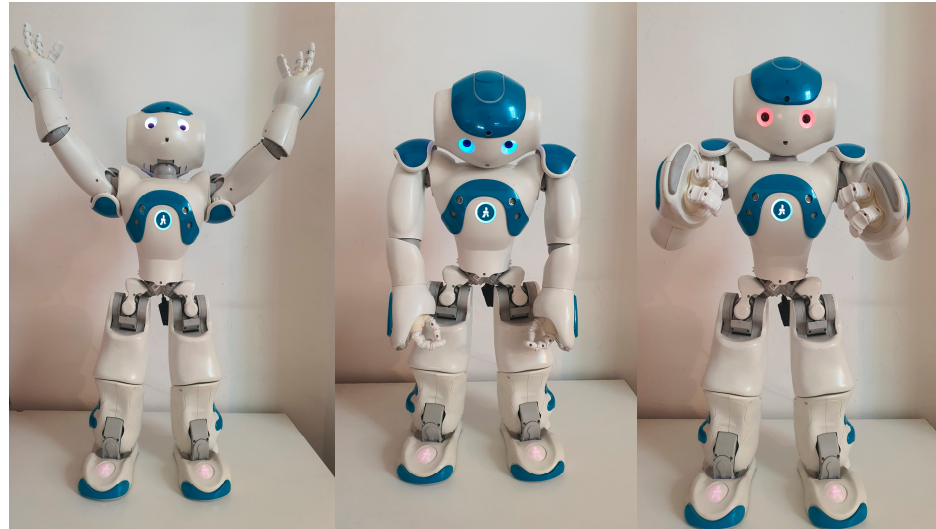
In the first session, the participants were asked to sit down in front of a computer screen together with NAO and to answer questions on demographic information. Then, they filled out the EmQue-CA [36].

In the second session, the children watched the three video clips in random order together with NAO (Figure 5). Each clip lasted for approximately 1 min each, and showed the emotions happiness, sadness, and anger. The clip that displayed the emotion happiness showed the final scene from the movie *Pinocchio*, where the puppet transforms into a real child and the cartoon characters express happiness and celebrate together. The clip that displayed the emotion sadness showed a clip of a girl (from the movie *Lilo and Stitch*) crying while holding her teddy bear. The clip that displayed the emotion anger, showed a section from the movie *The Incredibles*, where members of a family scream and yell at each other.



**Figure 5.** (a,b) The children watching video clips with NAO. The experimental setting also involved a screen for watching the video clips and a low table that allowed the child to sit near NAO.

After each clip, NAO responded to the movie with a congruent or incongruent emotion depending on the conditional condition. After watching a video clip and NAO's reaction, the children were asked to recognize and express the feelings displayed in the video and by the robot (see Figure 6).



**Figure 6.** Example of the emotions expressed by NAO (happiness, sadness, and anger).

In the third session, the children filled the items adapted by Tsiourti et al. [17] on the perceived credibility of the robot and the State Empathy Scale [39]. When the experiment was finished, the children had a small debriefing session, where they could interact and play with the robot and ask questions about the experiment.

### 3.5. Statistical Analysis

All statistical analyses were performed in SPSS for Windows (version 29). Basic descriptive statistics including means, standard deviations, percentages, frequency distributions, and Cronbach alpha for each scale were calculated. The items assessing cognitive empathy from the State Empathy Scale were removed from further analyses due to low Cronbach alpha values. The number of correct recognitions of NAO's emotions and emotions displayed in the video was summed. As the participants were asked to recognize three different emotions, the answers were distributed from 0 to 3, reflecting how many correct answers the children had. Initially, a correlation analysis between the different variables was conducted. A linear regression was conducted to predict emotion recognition in NAO. Differences between groups were examined with an independent t-test. A significance level (*p*-value) of less than 0.05 was applied for all tests.

## 4. Results

### 4.1. Descriptives

All means, standard deviations, and correlations between the variables are presented in Table 2. Emotion recognition in NAO and emotion recognition in the videos showed a correlation ( $r = 0.63$ ). Further, sex and emotion recognition by NAO showed a correlation ( $r = 0.54$ ).



**Table 2.** Descriptive statistics and correlations.

	M (SD)	1	2	3	4	5	6	7	8	9	10	11
1 Emotion recognition (NAO)	1.96 (1.09)	--										
2 Emotion recognition (video)	2.00 (0.78)	0.639 **	--									
3 Level of congruence <sup>1</sup>	1.44 (0.51)	-0.039	-0.097	--								
4 Hearing level <sup>2</sup>	1.33 (0.48)	-0.049	-0.102	0.00	--							
5 Empathy	2.70 (0.54)	-0.019	-0.181	-0.203	-0.049	--						
6 Affective empathy	3.59 (1.09)	0.148	0.325	0.044	0.104	-0.033	--					
7 Associative empathy	3.27 (1.30)	0.102	0.265	0.148	-0.072	0.131	0.815 **	--				
8 State empathy *	3.62 (0.89)	0.174	0.346	0.037	-0.043	0.082	0.935 **	0.915 **	--			
9 Believability	3.84 (0.84)	0.168	0.151	-0.451 *	-0.165	0.293	0.222	0.428 *	0.389 *	--		
10 Age	7.59 (1.50)	-0.033	0.196	0.045	0.569 **	.0130	0.030	0.009	0.037	-0.102	--	
11 Sex <sup>3</sup>	1.59 (0.50)	0.534 **	0.196	-0.017	-0.053	0.105	0.177	-0.003	0.133	-0.056	-0.178	--

Note. \*  $p < 0.05$ ; \*\*  $p < 0.01$ . <sup>1</sup> Level of congruence: 1 = congruent, 2 = incongruent. <sup>2</sup> Hearing level: 0 = hearing children, 1 = deaf children. <sup>3</sup> Sex: 1 = boy, 2 = girl.

#### 4.2. Credibility toward NAO

Overall, the children experienced NAO as credible ( $M = 3.84$ ;  $SD = 0.83$ ). The level of congruence predicted perceived credibility of the robot in a linear regression analysis ( $\beta = -0.451$ ,  $p = 0.01$ ), indicating that the children who saw NAO react to emotions in a congruent manner perceived the robot as more credible. The children in the congruent condition found it easier to imagine what the robot was thinking ( $M = 4.27$ ;  $SD = 1.22$ ) compared to the incongruent group ( $M = 3.0$ ;  $SD = 1.41$ ;  $t(25) = 2.496$ ,  $p = 0.02$ ). The children in the congruent condition also found NAO to respond more appropriately to the emotions in the videos, compared to the incongruent group ( $M_{\text{congruent}} = 4.73$ ,  $SD_{\text{congruent}} = 0.46$ ;  $M_{\text{incongruent}} = 2.75$ ,  $SD_{\text{incongruent}} = 1.71$ ;  $t(12.26) = 3.903$ ,  $p = 0.002$ ; see Table 3 and Table 4 for frequency distribution of responses to each item by the congruent and incongruent groups, respectively).

**Table 3.** Answers from the children in the congruent condition ( $N = 15$ ) on the questionnaire on perceived credibility of NAO, adapted by Tsiourti [17].

	Min	Max	M (SD)	To a Minor Extent	To Some Extent	To a Great Extent
The robot perceived the content of the movie clip correctly	2	5	4.53 (0.84)	6.7%	0%	93.3%
It was easy to understand which emotion was expressed by the robot	3	5	4.60 (0.63)	0%	6.7%	93.3%
It was easy to understand what the robot was thinking about	1	5	4.27 (1.22)	13.3%	0%	86.7%
The robot has a personality	1	5	3.80 (1.52)	20.0%	0%	80.0%
The robot's behavior drew my attention	1	5	4.00 (1.46)	26.7%	0%	73.3%
The robot's behavior was predictable	1	5	3.27 (1.75)	40.0%	6.7%	53.3%
The behavior expressed by the robot was appropriate for the content of the movie	4	5	4.73 (0.45)	0%	0%	100%

**Table 4.** Answers from the children in the incongruent condition (N = 12) on the questionnaire on perceived credibility of NAO, adapted by Tsiourti [17].

	Min	Max	M (SD)	To a Minor Extent	To Some Extent	To a Great Extent
The robot perceived the content of the movie clip correctly	1	5	3.83 (1.40)	25.0%	0%	75.0%
It was easy to understand which emotion was expressed by the robot	1	5	3.67 (1.44)	16.7%	16.7%	66.6%
It was easy to understand what the robot was thinking about	1	5	3.00 (1.41)	41.6%	16.7%	41.7%
The robot has a personality	1	5	3.00 (1.86)	41.7%	8.3%	50.0%
The robot's behavior drew my attention	1	5	4.42 (1.65)	8.3%	0%	91.7%
The robot's behavior was predictable	1	5	3.33 (1.49)	33.4%	8.3%	58.3%
The behavior expressed by the robot was appropriate for the content of the movie	1	5	2.75 (1.71)	58.3%	0%	41.7%

#### 4.3. Recognition of NAO's Emotions

A large proportion (81.5%) of the children reported that it was easy to guess NAO's emotions (M = 4.19; SD = 1.15). The children in the congruent group rated it easier to guess NAO's emotions (M = 4.60, SD = 0.63); however, there was no significant difference compared to the incongruent group (M = 3.67, SD = 1.44;  $p > 0.05$ ). The children were able to recognize, on average, just under two emotions displayed by NAO (M = 1.96 (SD = 1.09)). The emotions happiness and sadness were the most frequently correctly chosen emotions (n = 18, 66.7% for both emotions), followed by the emotion anger (n = 17, 63.0%).

In the regression analysis conducted to predict correct emotion recognition in NAO, only the ability to recognize emotions in the video ( $\beta = 0.720$ ,  $p < 0.001$ ) and sex ( $\beta = 0.421$ ,  $p = 0.01$ ) were found to be significant predictors (see Table 5 for a summary of regression results), indicating that girls were more likely to guess NAO's emotion correctly. The level of empathy and hearing level did not significantly predict the recognized emotions in NAO without sounds.

**Table 5.** Summary of regression results for predicting emotion recognition in NAO. All statistically significant values are emboldened.

Predictors	$\beta$	$t$	$p$
Emotion recognition (video)	0.720	4.341	<b>&lt;0.001</b>
Congruence level	0.134	0.742	0.469
Hearing level	0.166	0.794	0.439
Empathy	0.341	1.817	0.088
State empathy (affective)	0.107	0.182	0.858
State empathy (associative)	-0.073	-0.179	0.860
State empathy (total)	-0.400	-0.511	0.616
Credibility toward the robot	0.221	1.052	0.308
Age	-0.118	-0.579	0.571
Sex	0.421	2.757	<b>0.014</b>

Note:  $r^2 = 0.838$ .

## 5. Discussion and Conclusions

In this study, an experiment investigating the interaction between hearing and deaf children and a humanoid robot have been conducted. Further, in an experimental situation, the children witnessed the humanoid robot NAO responding with congruent or incongruent emotions to video clips, without using any sounds; that is, all emotions were expressed through non-verbal communications (i.e., with gestures and head movements). The children's ability to recognize the emotions expressed by NAO were investigated.

Overall, the children in this study reported medium to high credibility toward the robot. In line with findings from Tsiourti et al. [17], seeing NAO respond with congruent emotions to the video predicted higher perceived credibility of the robot. Further, the children who observed NAO's congruent emotions found the reactions to have been more appropriate and easier to recognize compared to the group that saw NAO's incongruent emotions. The findings indicate that it is important for children that the robot responds with emotions that suit a given situation for them to experience higher credibility toward it.

There was a varying ability to recognize emotions in NAO among the children in both groups. The emotions happiness and sadness were the most frequently correctly guessed emotions, which is in line with previous findings on emotion recognition in humans [40]. Even though several of the children reported that it was easy to recognize the emotions NAO exhibited, many also struggled to identify the right emotion expressed by NAO. Although the programming of emotions in NAO was inspired by previous studies [34,35], it is possible that the exhibited emotions were too indistinct for the children to recognize them correctly.

The ability to guess correct emotions displayed in the video clips predicted the ability to recognize the emotions NAO displayed, regardless of the congruence level of the emotion. Previous findings suggest that when NAO exhibits congruent emotions to video clips, the recognition of emotions in the robot are simplified [17]. Contrary to these previous findings, the children in this study who observed NAO respond with congruent emotions were not more likely to recognize the correct emotion, compared to the group that observed NAO respond with incongruent emotions. Several factors can explain these contradictory results. First, factors such as general self-concept and age have been found to influence emotion recognition in humans [40]. This study's young sample may therefore be an explanation for why the previous results were not replicated in this experiment. Further, it is also important to mention that the small sample size may have been an influencing factor as to why previous findings were not replicated in this study.

There was no difference between hearing and deaf children's ability to recognize emotions in NAO. Even though no difference was found in this study, previous findings have suggested that deaf children have a poorer ability to recognize emotions in others compared to hearing children [17,40]. Wiefferink et al. [41] showed that children with cochlear implants recognized fewer emotions than children with normal hearing and that emotion recognition skills are affected by hearing loss, including when using non-verbal communication to express emotions. Finally, even though empathy has been shown to influence experiences in human–robot interactions [27], the results of this study showed, surprisingly, that empathy did not lead to a better ability to recognize emotions in the robot. No difference in empathy was found between the conditions or groups, in contrast to previous findings suggesting that deaf children have lower levels of empathy than hearing children [20–22].

This study aims to contribute to the future research on the field of human–robot interactions between deaf children and robots. As well as to contribute to a broader understanding of how children recognize emotions exhibited from humanoid robots.

However, several limitations need to be considered when interpreting the presented results. First, the sample size is relatively small, which affects how generalizable the findings are to a broader population. In addition, the study focused only on the three emotions: happiness, sadness, and anger. This is a potentially limiting factor for the exploration of children's ability to recognize more comprehensive emotions in the robot. In future research,

all primary emotions should be included to gain a broader understanding of children's emotion recognition in robots. Furthermore, the study mainly involved a sample with similar cultural backgrounds, potentially limiting the generalizability of the results to a more diverse population. Equally, to avoid a too indistinct display of emotions by the robot with non-verbal communication, future research should investigate how to program robots to express properly intended emotions. This study examined deaf and hearing children's interactions with a humanoid robot. The children reported, overall, a medium-to-high level of credibility toward the robot.

Further, both hearing and deaf children observed a robot watching video clips that displayed the emotions happiness, sadness, and anger. The robot displayed either a congruent or incongruent emotion in response to the video clips, and the children's ability to recognize the emotions in the robot was examined. No differences were found between the congruent and incongruent conditions. Notably, the ability to recognize emotions in a video and sex were identified as significant to identify emotions in the robot. Future research should include larger samples, incorporate diverse cultural perspectives, and encompass a broader range of emotions to enhance the generalizability and depth of understanding in this field.

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**Informed Consent Statement:** Informed consent was obtained from all parents of the participating children involved in the study.

**Data Availability Statement:** The data presented in this study are available on request from the corresponding author. The data are not publicly available due to them containing information that could compromise research participant privacy.

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