

Brain activity during reciprocal social interaction investigated using conversational robots as control condition

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2	Brain activity during reciprocal social interaction
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5	Birgit Rauchbauer ^{1, 2, 4} , Bruno Nazarian ¹ , Morgane Bourhis ¹ ,
6	Magalie Ochs ³ , Laurent Prévot ^{4, 5} & Thierry Chaminade ¹
7	
8	1. Institut de Neurosciences de la Timone, Aix-Marseille Université, CNRS, France
9	2. Laboratoire de Neurosciences Cognitive, Aix-Marseille Université, CNRS, France
10	3. Laboratoire d'Informatique et des Systèmes, Aix-Marseille Université, CNRS, France
11	4. Laboratoire Parole et Langage, Aix-Marseille Université, CNRS, France
12	5. Institut Universitaire de France, Paris France
13	
14	Abstract
15	We present a novel functional magnetic resonance imaging (fMRI) paradigm for second-person
16	neuroscience. The paradigm compares a human social interaction (human-human interaction,
17	HHI) to an interaction with a conversational robot (human-robot interaction, HRI). The social
18	interaction consists of 1-minute blocks of live bidirectional discussion between the scanned
19	participant and the human or robot agent. A final sample of 21 participants is included in the corpus
20	comprising physiological (BOLD, respiration and peripheral blood flow) and behavioural
21	(recorded speech from all interlocutors, eye tracking from scanned participant, face recording of
22	the human and robot agents) data. Here we present the first analysis of this corpus, contrasting
23	neural activity between HHI and HRI. We hypothesized that, independently of differences in
24	behaviour between interactions with the human and robot agent, neural markers of mentalizing
25	(temporo-parietal junction and medial-prefrontal cortex) and social motivation (hypothalamus and
26	amygdala) would only be active in HHI. Results confirmed significantly increased response
27	associated with HHI in the temporo-parietal junction, hypothalamus and amygdala, but not in the
28	medial prefrontal cortex. Future analysis of this corpus will include fine-grained characterization
29	of verbal and non-verbal behaviours recorded during the interaction to investigate their neural
30	correlates.
31	Introduction

32 Humans' social bonds are established and maintained through interactions with others. These interactions "are characterized by intricate reciprocal relations with the perception of 33 socially relevant information prompting (re-) actions, which are themselves processed and reacted 34 to" (Schilbach et al., 2013). To date, the field of social neuroscience, investigating the 35 neurophysiological basis of social interactions, has mostly focused on the investigation of the 36 observation of social signals, rather than on truly interactive social settings. In an attempt to 37 capture the interactional dynamics in real-life, "second person neuroscience" (Schilbach, 2015; 38 Schilbach et al., 2013) encourages the investigation of naturalistic interactive paradigms for 39 enhanced ecological validity (Pan & Hamilton, 2018). This approach aims to shift social 40 neuroscience from a prevailing "passive spectator science" (Hari, Henriksson, Malinen, & 41 Parkkonen, 2015) to an approach investigating the dynamics of social exchange (Hari et al., 2015). 42 This requires that not only the experimental, but also control condition should preserve the 43 44 reciprocity of real-time interactions.

Computer-animated on-screen agents have been used to study the influence of animacy on 45 motor imitation (Klapper, Ramsey, Wigboldus, & Cross, 2014) and mechanisms of joint attention 46 (Schilbach et al., 2006, 2010; Wilms et al., 2010), taking advantage of the extensive control the 47 48 experimenter can have on their behaviour. This includes for example control over the direction of the gaze, towards or away from a target, or its timing. Robots also have been used as control 49 conditions in social neuroscience experiments (Chaminade et al., 2010, 2012; Chaminade, Da 50 51 Fonseca, Rosset, Cheng, & Deruelle, 2015; Krach et al., 2008). This article presents a novel 52 second-person neuroscience paradigm for functional magnetic resonance imaging (fMRI) that uses a conversational robot as control condition for a human social interaction (Hale et al., 2018)... 53 Social interaction is operationalized using language, the most ubiquitous form of human 54 interaction. The paradigm allows recording brain activity during 1-minute live bidirectional 55 discussions between the scanned participant and a fellow human (human-human interaction; HHI) 56 and similar discussions between the same participant and a conversational robot (human-robot 57 interaction; HRI). 58

The HHI represents the experimental condition, constituting the "social" condition. The HRI represents the control condition, which preserves sensorimotor aspects of live, bidirectional conversation. Indeed, the robot has an anthropomorphic outer appearance, including a human face and voice, so that seeing, hearing and talking to the artificial agent is similar to the interaction with the human agent. In addition, while participants believe the robot is autonomous, it is actually 64 controlled by the same individual participants discuss with in the HHI conditions. As a 65 consequence, participants are not aware that they interact with the same individual, the 66 confederate, in both HHI and HRI conditions (see Figure 1, top). On the other hand, the 67 conversational robot used in the experiment is clearly not human: the face is projected on a 68 moulded plastic screen, it has a limited number of pre-scripted sentences for conversation and it 69 doesn't exhibit meaningful facial expressions or speech intonations.

A corpus of multimodal data is collected in addition to the fMRI data. Physiological 70 responses (respiration and peripheral blood flow pulse) are recorded synchronized with the MR 71 scanner and are used, currently, for modelling and removing physiological noise in the fMRI data. 72 Behavioural data is recorded to enable future exploration of brain-behaviour relations. This 73 includes speech production by the scanned participants and human and robot agent, the video 74 capture of the human and robot agent, and the gaze movement of the scanned participant. Given 75 76 the unconstrained nature of the conversation task, a fine-grained exploration of the behaviour, in 77 particular transcription and analyses of conversations, and exploration of dynamic gaze direction 78 to the human and robot agents' face, will be necessary to explore brain correlates in the corpus. Here, we focus on the block analysis by contrasting conditions HHI and HRI. Given that the robot 79 80 control condition is designed to reproduce sensorimotor aspects of human conversation, both HHI 81 and HRI are expected to be associated with a neural network involved in visuomotor speech perception and in speech production, including bilaterally the dorsal temporal lobes for speech 82 83 perception the ventral and lateral occipital cortex for face perception, as well as the bilateral ventral primary motor cortex (speech motor control) and the left inferior frontal gyrus ("Broca's area") 84 for speech production (see Price, 2012 for review). 85

86 The contrast between conditions HHI and HRI is used to test specific hypotheses about the neural correlates of social cognition, and hence confirming the quality and validity of the acquired 87 data. Social cognition (Adolphs, 1999) is broadly defined as "the sum of those processes that allow 88 individuals of the same species (conspecifics) to interact with one another." (Frith & Frith, 2007). 89 On the basis of previous work (Chaminade et al., 2010, 2012, 2015; Gallagher, Jack, Roepstorff, 90 & Frith, 2002; Krach et al., 2008), we specifically expected processes of mentalizing and enhanced 91 92 social motivation when interacting with the human compared to the robot. Mentalizing is the ascription of mental states such as intentions and beliefs to explain the apparent behaviour of the 93 interaction partner (Frith & Frith, 1999). It requires the adoption of an intentional stance towards 94 95 the interaction partner – the assumption that the interacting agent actually has a mind supporting

96 mental states (Dennett, 1989). The adoption of an intentional stance towards a human versus a computer interaction partner has been linked to activation in the paracingulate cortex (Gallagher 97 et al., 2002), a region of the medial prefrontal cortex (MPFC). It has been argued that humans do 98 not adopt an intentional stance towards robots, computers and more generally artificial agents 99 100 (Dennett, 1989). Indeed, increased activity in areas associated with mentalizing, not only in the medial prefrontal cortex (MPFC), but also the temporoparietal junction (TPJ) has been repeatedly 101 102 found when interacting with a human compared to a robot or a computer (Chaminade et al., 2010, 2012, 2015; Gallagher et al., 2002; Krach et al., 2008). Such neural markers of mentalizing are 103 expected in the contrast HHI versus HRI. 104

105 Also on the basis of previous results in experiment contrasting human versus robot interactions (e.g. Chaminade et al., 2012; Chaminade et al., 2015) we expected human interaction 106 to elicit activation of neural markers of social motivation, the human drive to interact, establish 107 108 and maintain bonds (Chevallier, Kohls, Troiani, Brodkin, & Schultz, 2012). Chaminade and colleagues (2015) report that a modulation of activity located in the paraventricular nucleus of the 109 hypothalamus by the social context (human versus robot) is present in neurotypical but not in 110 individuals diagnosed with autism spectrum disorder. This was associated with the proposal that 111 autism is associated with a deficit in social motivation, involving disrupted hypothalamic 112 113 regulation of oxytocin release (Chevallier et al., 2012). Consecutive works confirmed the modulation of hypothalamus anatomy (Wolfe, Auzias, Deruelle, & Chaminade, 2015) and activity 114 115 (Wolfe, Deruelle, & Chaminade, 2018) by the social context. In general, social motivation and reward have been associated with brain activation in the reward-circuit, comprising the ventral 116 striatum, orbitofrontal and ventromedial cortex (Chevallier et al., 2012), including amygdala 117 specifically for social reward (Rademacher et al., 2010). In line with these studies, we expected 118 that the interaction with a human would activate the previously reported subcortical areas (in 119 particular hypothalamus and amygdala) more than interaction with a robot. In contrast, we had no 120 specific hypothesis with regards to brain activity in the reverse contrast HRI versus HHI. 121

In the next sections we present the experimental paradigm, and the first results of the reciprocal contrasts between conditions HHI and HRI, demonstrating not only the feasibility of our approach, but also the scientific quality of the acquired data with regards to our hypotheses.

125

126 Methods

127 Participants

Twenty-four native French-speaking participants (7 men) with an average age of 28.5 128 (SD=12.4) were fMRI scanned while having a conversation with a fellow human or a 129 retroprojected conversational robotic head (Furhat robotics, https://www.furhatrobotics.com/; Al 130 Moubayed, Beskow, Skantze, & Granström, 2012). Three participants were excluded due to 131 132 technical problems and insufficient task compliance. Twenty-one participants (mean age = 25.81, SD = 7.49) were included in the analysis. Participants received information about the experiment, 133 confirmed their compatibility for MR scanning and gave their informed consent prior to scanning. 134 Eligibility entailed normal or corrected-to-normal vision and no history of psychiatric or 135 neurological conditions. Participants received a flat fee of 40 Euro for participation. The study was 136 approved by the ethics committee "Comité de Protection des Personnes Sud Méditerrannée I". 137

138

139 Cover story for the experiment

A recent behavioural study comparing human-human with human-robot conversations 140 (Chaminade, 2017) was adapted to the fMRI environment. The experimental factor was the nature 141 of the INTERACTING AGENT (HUMAN versus ROBOT), in a within-subject, block-design. A cover 142 story was a fundamental element of the study, as it provided a fake rationale for the experiment as 143 144 well as a frame for discussion and explanations for the experimental set-up. Volunteers were told 145 they participated in a neuromarketing experiment sponsored by an advertising company. The company wanted to test whether the message of their forthcoming advertisement campaign can be 146 147 identified if a pair people are presented the images of the campaign and discuss about them. Two series of three images presented anthropomorphized fruits and vegetables as superheroes or 148 appearing rotten respectively (see Supplementary Material, Figure S1). Participants were 149 instructed to talk freely about the presented image with the agent outside the scanner, either a 150 human or the conversational robotic head (controlled by the confederate, unbeknown to the 151 participant; see next section). The robot was a presented as an autonomous conversational agent 152 that had information about the advertisement campaign. As such, the discussion with the robot 153 could be used to gather information about the advertisement campaign. 154

In practice, the cover story was presented to the participants by experimenter BR in the lobby of the MR centre, later joined by the confederate. Confederates were gender-matched to participants. Experimenter TC served as confederate for men and experimenter MB for women. The participant was told that the confederate had already participated in the experiment inside the scanner and had agreed to come back to play the role of the agent outside of the scanner. The participant was then accompanied into the control room outside the scanner and shown the robot (see next section). In the meantime, we asked the confederate to wait, telling him/her that we would first get the participant ready into the scanner. At the end of the experiment, participants were debriefed verbally to verify they still believed in the cover story and we revealed the true objective of the experiment.

165

166 Artificial agent

The robotic head from Furhat robotics (https://www.furhatrobotics.com/; Al Moubayed, 167 Beskow, Skantze, & Granström, 2012) was used in this study. The robotic head is a semi-168 transparent plastic mask moulded to mirror the shape of a human face on which the image of a 169 human face is retro projected. In order to match the robot appearance to the confederates, the face 170 and voice were gender-matched, a wig, a scarf and headphones were added as well as glasses for 171 confederate TC (see illustrations in Figure 1). Furhat OS allowed us to control its responses 172 through a Wizard of Oz (WOZ): unbeknown to the participant, the confederate was controlling the 173 robot remotely. The robot conversational feedbacks were largely based on actual human 174 interactions recorded during the previous the behavioural study (Chaminade, 2017). A WOZ user 175 interface was created with Furhat OS displaying buttons on a web browser running on a tablet 176 allowing the human controller to launch pre-programmed conversational feedback. For example, 177 clicking the button "yes" on the screen would make the robot say "yes", clicking the button 178 179 "superhero" would launch the sentence "It looks like a superhero". Conversational elements included non-specific feedbacks, such as "yes", "no", or "maybe", which could be used for all 180 images, as well as specific feedbacks for each of the images, such as "This lemon looks like a 181 superhero" or "Maybe this is a campaign to eat healthier food". Note that the cover story allowed 182 to limit the number of targeted conversations for each image compared to unconstrained 183 discussion. Overall about 30 French conversational feedbacks were scripted for the robot for each 184 of the six images (see Supplementary Material, file S1 for robot statements). 185

The robot was controlled using this WOZ interface by the confederate acting as conversational agent in the human condition, allowing for a realistic bidirectional conversation similar to the interaction with the human. Thus, unbeknown to the participants, they discussed with the same agent in both human and robot conditions. On the other hand, while the conversational robot was able to reproduce superficial aspects of a human conversation, it lacked intonations in speech, head movements, facial expressions and the ability to elaborate longer statements, thus 192 appearing clearly artificial and participants believed, according to debriefing, that it was 193 autonomous.

194

195 Experimental set-up

196 The fMRI audio set-up allowed live conversation between the scanned participant lying supine in the scanner and the agent outside of the scanner despite the noisy MRI environment. It 197 consisted of an active noise-cancelling MR compatible microphone (FORMI-III+ from 198 optoacoustics mounted on the head coil) and insert earphones from Sensimetrics. Live video of 199 200 the interacting agent (human or robot) was captured by webcams and projected to a mirror mounted on the antenna in front of the scanned participant's eyes. Videos were recorded for future analysis. 201 Participants' direction of gaze on the projection mirror was recorded (Eyelink 1000 system, SR 202 Research). Stimulus presentation, audio and video routing and recording, synchronization with the 203 fMRI acquisition triggers and the eye tracker was implemented in a Labview (National Instrument) 204 virtual machine (see Figure 1, top). Finally, blood pulse and respiration were recorded with built-205 in Prisma Siemens hardware and data format. 206

Altogether, we collected multimodal data including behaviour (speech from the participant 207 208 and human or robot agent, video capture of the human and robot agent, and the gaze movement of 209 the scanned participant) and physiology (BOLD signal, respiration and peripheral blood flow pulse) to form a corpus. Transcribed speech data (more details on the transcription and an example 210 211 of the conversation is provided as Supplementary Material in section 2.1 and Files S3-S5) and 212 fMRI data, both raw and analysed, will be shared in online repositories.

213

Experimental paradigm 214

The MRI recordings consisted of four sessions of each six 1-minute blocks of conversation 215 each, showing the "super-heroes" images in the first and third sessions and the "rotten fruits" 216 images in the second and fourth sessions (see Supplementary Material, Figure S1 and Table S1 for 217 details). The order was kept constant across participants each session alternating the three images 218 per session and two INTERACTING AGENTS (complete order of conditions is given in Supplementary 219 220 Material Table S1). Each image was thus shown twice in each session, once per INTERACTING AGENT. Given the entertaining nature of the interaction, we did not expect habituation effects to 221 affect the brain imaging data and preferred to have the nature of the agent fully predictable. Hence, 222 223 we did not randomize the order of presentation of the human and robot agents.

224 Blocks started with the presentation of one image for 8.3 seconds, followed by a 3.3 second black screen, after which there was a live bidirectional conversation with the INTERACTING AGENT 225 for one minute, followed by an inter block interval black screen of 4.6 seconds (figure 1, bottom). 226 In the absence of live video feed from inside the scanner, a light signalled to the confederate that 227 228 the conversation had started. The participant initiated the conversation, instructed to talk freely with the other agent about the image and their suggestions on the topic of the advertisement 229 campaign. One block lasted 76.2 seconds and one session 8 minutes and 2 seconds of fMRI 230 recording. We recorded 3 minutes of conversation per INTERACTING AGENT and session, for a total 231 of 24 minutes of conversation per participant. Audio and video set-up of the conversation was 232 tested beforehand, and audio adjusted individually for each participant. As participants were 233 always connected via audio with the confederate, some indicated that the sound level wasn't 234 appropriate thus giving us the chance to adapt the audio if required. This information was recorded 235 236 for future use.

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MRI acquisition

241 MRI data was collected with a 3T Siemens Prisma (Siemens Medical, Erlangen, Germany) using a 20-channel head coil. Blood oxygen level-dependent (BOLD) sensitive functional images 242 243 were acquired using an EPI sequence in the 4 runs. Parameters were as follows: Echo time (TE) 30 ms, repetition time (TR) 1205 ms, flip angle 65°, 54 axial slices co-planar to the anterior / 244 posterior commissure plane, FOV 210mm x 210mm, matrix seize 84 x 84, voxel size 2.5 x 2.5 x 245 2.5 mm³, with multiband acquisition factor 3. After functional scanning, structural images were 246 acquired with a GR IR sequence (TE/TR 0.00228/2.4 ms, 320 sagittal slices, voxel size 0.8 x 0.8 247 x 0.8 mm³, field of view 204,8 x 256 x 256mm). 248

- Figure 1 around here -

249

250 MRI data analysis

MRI using SPM12 Parametric Mapping, 251 data analysed (Statistical was 252 http://www.fil.ion.ucl.ac.uk/spm/). First, we calculated the voxel displacement map. The time series for each voxel was then realigned temporally to the acquisition of the slice in the middle in 253 time to correct for differences in slice time acquisition. The image time series were unwarped using 254 255 the voxel-displacement map to take into account local distortion of the magnetic field and spatially 256 realigned using a sinc interpolation algorithm that estimates rigid body transformations (translations, rotations). Images were then spatially smoothed using an isotropic 5 mm full-width-257 at-half-maximum Gaussian kernel. The first realigned and unwarped functional image was 258 coregistered with an unwarped single-band reference image recorded at the onset of each trial, 259 260 which was itself coregistered with the T1 and T2 anatomical images. These anatomical images were segmented into grey matter (GM), white matter (WM), and cerebral spinal fluid (CSF) using 261 SPM12 "New segment". GM, WM, and CSF tissue probability maps from our sample of 21 262 included participants were used to form a DARTEL template (Ashburner, 2007). The deformation 263 flow fields from individual spaces to this template were used to normalize the beta images resulting 264 from the individual subjects' analyses (i.e. in subjects' individual space) for use in a random-effect 265 second-level analysis. 266

Potential artefacts from blood pulse and respiration were controlled using the Translational Algorithms for Psychiatry-Advancing Science (TAPAS) toolbox standard procedure (https://www.tnu.ethz.ch/de/software/tapas/documentations/physio-toolbox.html; Kasper et al., 2017). Realignment parameters (translation and rotation) as well as their derivatives and the square product of both parameters and their derivatives were used as covariates to control for movementrelated artefacts. We also used the Artefact Detection Tools (ART) to control for any movementrelated artefacts (<u>www.nitrc.org/projects/artifact_detect/</u>) using the standard threshold of 2 mm.

The fMRI time series were analysed using the General Linear Model (GLM) approach implemented in SPM. Single-subject models consisted of one regressor representing the oneminute discussion for each of the two INTERACTING AGENTS, and another one representing the presentation of the images.

After normalization, beta estimates images were entered in a mixed-model analysis of variance (using SPM "full ANOVA") with participants and sessions as random factors and the nature of the INTERACTING AGENT as factor of interest for inferences at the population level. A mask was created on the basis of the mean of DARTEL normalized anatomical GM and WM tissue classes of each participant, also used for rendering results in Figures 2 & 3.

We first assessed the main effect of the conversation with both agents against the implicit baseline. We then looked specifically at the effects of each of the INTERACTING AGENT contrasted to the other one, with a clear focus on brain areas involved in mentalizing and social motivation in the contrast HHI *versus* HRI. All statistical inference was performed applying a threshold of p = 0.05 False-Discovery Rate (FDR) corrected for the whole brain at the cluster-level (Friston, Holmes, Poline, Price, & Frith, 1996). Anatomical localization of the resulting clusters relied on the projection of the results onto the mean anatomical image of our pool participants resulting from DARTEL coregistration.

291

292 Results

293 *Cover story debriefing*

A verbal debriefing was performed in an undirected and open format, to allow the participants to report their experience in an unbiased manner. None of the participants reported feelings of distress during the experiment with either interaction partner, or doubts about the autonomous nature of the conversational robot. In conclusion, all participants still believed in the cover-story at the end of the recordings.

299

300 Assessment of participants' movements during scanning

No participant was excluded on the basis of the assessment of movement using the toolbox ART (https://www.nitrc.org/projects/artifact_detect/). At the movement threshold used, between 0 and a maximum of 3 volumes per session and participant were considered as outliers. In the absence of large artefacts, all scans and sessions from the twenty-one participants were included in the analysis. Moreover, using the same metric to calculate a global movement per block of discussion, session and subject, an analysis of variance showed no effect of the INTERACTING AGENTS (F(1,495)=2.22, p = 0.14; see Supplementary Figure S2).

308

309 Participants' behaviour

The full transcription of the 504 minutes of discussion collected for the corpus is ongoing (examples, as well as the link to the data repository are presented in supplementary material; see Files S3-S5). Yet is has been observed by the confederates that discussions between the two agents differed in terms of the speed and emotion conveyed by participant's voice. Participants spoke in general faster and with increased prosodic variations with the human than the robot agent. Humour was also observed in the conversation with the human, but not with the robot. These observations are expected given the differences in conversational competence between the two agents.

317

318 *fMRI results*

319 The main effect of conversation for the human and for the robot largely overlapped (Figure 2, top). As predicted given the nature of the task, common activation clusters are found bilaterally 320 along the superior temporal sulcus and gyrus, the central operculum, the lateral and ventral 321 occipital cortex, the lateral premotor cortex, the supplementary motor area and the ventral and 322 323 dorsal cerebellum, as well as the left inferior frontal gyrus. Differences between the resulting activation maps for the human and robot agents were quantitative rather than qualitative, with 324 325 larger clusters mostly related to motor control (in region of the precentral and postcentral gyri) for the robot and to speech processing (in the temporal cortex) for the human. 326

The contrast HHI versus HRI (see Figure 2, bottom left) revealed bilateral activation in the 327 superior temporal gyrus and sulcus that overlapped partly with the temporal areas associated with 328 the main effects of the conversations. It extended anteriorly to the temporal poles and to the 329 posterior lateral orbitofrontal cortex. Posteriorly, it covered the temporo-parietal junction and 330 331 lateral occipital cortex. Another significant cluster covered a number of subcortical structures: the bilateral thalamus, hypothalamus, hippocampus, amygdala, caudate nucleus and the subthalamic 332 area. We also found bilateral activation in the cerebellum centred on the horizontal fissure. No 333 334 medial prefrontal cluster was found at the threshold used.

The reverse contrast HRI *versus* HHI identified a number of bilateral activation clusters. In the occipital region, a cluster centred on the striate cortex extended to the lingual and fusiform gyri. Furthermore, a strong activation was found bilaterally within the intraparietal sulcus extending to the supramarginal gyrus. Clusters were also found in the middle frontal gyrus and the centred on the lateral central sulcus.

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- 341

- Figure 2 and 3 around here -

342 **Discussion**

We introduce a novel paradigm to investigate the neural bases of natural interactions between humans, in line with a second-person neuroscience approach. We choose live bidirectional conversation as operationalization of natural interactions given it is the most common form of communication between humans. The scientific challenge is twofold.

Methodologically, investigating natural interactions implies that the classical experimental approach, in which only one parameter is changed between experimental conditions, isn't applicable. Here, we use a robot for a high-level control condition: the conversational robot reproduces a number of sensorimotor aspects of the conversation, yet is far from mimicking a real human, and it does not elicit the adoption of an intentional stance according to Dennett (1989).
Hence, interacting with the robot in the current paradigm can be considered to be non social,
yielding a unique control condition for the social interaction with a fellow human.

Technically, the constraints of MRI recordings are numerous for a live bidirectional conversation during fMRI scanning: participants lie supine in a very noisy environment and are required to avoid any movement to ensure the quality of the data. We decided to hold the head firmly using foam pads while keeping the jaw free. Importantly, post-hoc assessment of individual participants' movements showed very limited motion and no quantitative difference between the human and robot condition, confirming the feasibility of the task.

The main objective of the analysis presented in the present article is to evaluate the quality of the recorded fMRI data, the main part of a unique corpus of neural, physiological and behavioural data. We have strong hypotheses about brain responses expected to be common during conversation with the two agents, as well as for the difference between interaction with the human *versus* the robot.

365

Commonly activated areas. We report a large number of common activated areas in the main 366 367 effects of HHI and HRI that can be directly related to sensorimotor aspects of the conversation. As 368 expected, they cover the dorsal half of the posterior temporal cortex bilaterally, known as the main brain region for auditory speech perception, comprising functional areas such as the primary 369 370 auditory cortex or temporal voice areas (Belin, Fecteau, & Bedard, 2004; Belin, Zatorre, Lafaille, 371 Ahad, & Pike, 2000). Common activations are also found in motor-related areas which are involved in the motor aspects of speech production. In particular, the ventral and opercular region 372 373 below the central suclus and adjacent precentral and postcentral gyrii is likely to include primary motor and sensory regions involved in verbalization (e.g., Price, 2012), while the lateral cluster in 374 the central sulcus area maps into the sensorimotor representation of the larynx (Brown et al., 2009). 375 The lateralized inferior frontal gyrus corresponds to Broca's area, crucial for the production of 376 377 speech. The medial premotor areas and the cerebellum are generally associated with the timing of action, which is crucial for articulation (see for review Price, 2012). Note that these motor areas 378 379 could also be involved in speech perception according to the motor theory of speech perception (Galantucci, Fowler, & Turvey, 2006). Indeed, a recent study revealed correlated 380 activation in the temporal auditory areas and the inferior frontal gyrus during successful coupling 381 382 between a speaker and a listener during a delayed interaction (Stephens, Silbert, & Hasson, 2010).

Current results show that live bidirectional conversation, irrespective of the agent, activates a network of brain regions previously associated with speech perception and production. Unfortunately, speech production and perception can't be distinguished in the current analysis, but will be the object of future exploration of this corpus. Finally, the large cluster spanning the lateral and ventral occipital cortex most likely responds to the processing of visual information, namely the face of the human or robot agent talking.

389

Increased activity areas in HHI condition. Interaction with a fellow human as compared to the 390 robot, revealed activation in the temporal cortex, including the bilateral temporoparietal junction 391 (TPJ), and subcortical activation in the hypothalamus, the thalamus, the hippocampus, the 392 amygdala and the subthalamic area. The results are in line with our predictions, except for the 393 absence of activation in the anterior medial frontal cortex. Activation in the TPJ and hypothalamus 394 395 has been reported in previous studies comparing human to robot interaction (Chaminade et al., 2012, 2015; Krach et al., 2008). TPJ activation has recently been reported when explicitly 396 ascribing human intention to robot behavior (Özdem et al., 2017). Hypothalamus activation during 397 HHI versus HRI was linked to enhanced social motivation (Chaminade et al., 2015; Chevallier et 398 399 al., 2012), given the release of oxytocin by hypothalamus subnuclei (Bartz et al., 2011; Heinrichs, von Dawans, & Domes, 2009). The amygdala has specifically been related to social as compared 400 to monetary reward (Rademacher et al., 2010). It is a key neural node in the processing of 401 402 emotionally and socially relevant information, coding saliency, reward and value of social stimuli 403 (Adolphs, 2010).

404

Increased activity areas in HRI condition. The contrast HRI versus HHI showed significant 405 activation in visual areas, including the fusiform cortex, hosting the fusiform face area, in the 406 intraparietal sulcus and in anterior parts of the middle frontal gyrus. We did not have specific 407 hypotheses for the effect of HRI compared to HHI, so all interpretation remains speculative. The 408 finding of enhanced activity in an area prominently involved in human face perception (FFA) has 409 been previously reported for action observation comparing robotic to human movements (Cross, 410 411 Ramsey, Liepelt, Prinz, & Hamilton, 2016), and, alongside enhanced activation in visual areas, for the perception of robot compared to human faces (Chaminade et al., 2010). This has been 412 interpreted as additional visual processing effort to identify an unfamiliar, robotic face (Chaminade 413 414 et al., 2010). Interestingly, the intraparietal sulcus was associated with the "uncanny valley" effect (Saygin, Chaminade, & Ishiguro, 2010), and interpreted as reflecting an increase of attention
towards unfamiliar stimuli. Enhanced response in visual areas, the IPS and the MFG seems in line
with studies investigating mechanistic versus social reasoning (Jack, Dawson, Begany, et al., 2013)
and ratings of images depicting machines (including robots) versus humans (Jack, Dawson, &
Norr, 2013).

Overall, we largely confirmed our hypotheses for brain activation in response to human compared to robot conversation. They support that processes of mentalizing and social motivation are enhanced in our paradigm when interacting with a human rather than with a robot. These results further confirm the quality and validity of the brain imaging data recorded, the main part of corpus also including behavioural and physiological data collected with the approach presented in this paper.

426

427 Limitations

We present an approach towards truly reciprocal, interactive social neuroscience and first supporting neurophysiological results. One major concern in fMRI studies involving language is the risk of extensive movement artefacts induced by motor-related aspects of speech-production. Yet, in the present study, we observed hardly any speech-induced movement during recording (see Supplementary Figure S2).

The pre-scripted sentences of the robot were shorter and more limited than the human's. The robot intonation, and more generally of head and face movements, were not controlled in the current experiment. Thus, it is expected that the human conversation differed from the robot conversation. This is likely to explain some of the differences in brain activity reported here.

The univariate fMRI analysis presented here is not sufficient to investigate the complex dynamics of the interactions. The corpus collected contained not only fMRI but also behavioural (linguistic, eye-tracking of the participant, video of the other agent during the interaction) and physiological (respiration and blood pulse) data. Future work on the corpus will entail fine-grained description of the behaviour, that will fuel the analysis of fMRI data. Transcription of speech recordings is under way (see supplementary information for an example of transcription) and will be made publicly available together with the fMRI data.

444 Also, future studies should include explicit measures of the perception of robots in general, 445 and of the conversational robot used in the experiment more specifically, in the form of questionnaires that would provide insights about individuals' variations in their expectations aboutthe robot's capacity.

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449 Conclusion

We investigated natural interaction comparing Human-Human Interaction (HHI) and 450 Human-Robot Interaction (HRI) using fMRI. Using a conversational robot as control condition 451 allowed to preserve reciprocal dynamics during interaction. Results for HHI showed activity in 452 brain areas associated with mentalizing and social motivation. The article introduces an innovative 453 454 paradigm in a second-person neuroscience approach. As such, it could be used as a starting point for social neuroscience to investigate specificities of human social cognition as well as to quantify, 455 and thus participate in the improvement of, the social competence of robots interacting with 456 humans. 457

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Figure 1:

Experimental design showing (a) the communication between the scanned participant and the other conversation agent, either the confederate or the robot, as well as the recording modalities; (b) the timeline of the experiment showing the alternation between the stimuli and conversation periods, as well as the relative timing. The fruit pictures correspond to the images used in the cover story, while the robot and confederate pictures illustrates episodes of live bidirectional conversations.

Figure 2:

Render of the brain surface of the mean of the coregistered and normalized brains from our participants sample. Overlaid are the results of the contrasts of interest (p < 0.05 FDR-corrected at the cluster level). Upper row shows the contrast of the human-human interaction (HHI) *versus* baseline in blue, and of the human-robot interaction (HRI) *versus* baseline in red. Lower row shows the contrast HHI *versus* HRI in blue, and HRI *versus* HHI in red.

Figure 3:

Coronal (top images), sagittal (middle images) and axial (bottom images) sections focusing on the cluster identfying subcortical structures significantly activated in HHI *versus* HRI.











