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The Control of Automatic Imitation Based on Bottom–Up and Top–Down Cues to Animacy: Insights from Brain and Behavior

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Abstract

Humans automatically imitate other people’s actions during social interactions, building rapport and social closeness in the process. Although the behavioral consequences and neural correlates of imitation have been studied extensively, little is known about the neural mechanisms that control imitative tendencies. For example, the degree to which an agent is perceived as human-like influences automatic imitation, but it is not known how perception of animacy influences brain circuits that control imitation. In the current fMRI study, we examined how the perception and belief of animacy influence the control of automatic imitation. Using an imitation–inhibition paradigm that involves suppressing the tendency to imitate an observed action, we manipulated both bottom-up (visual input) and top-down (belief) cues to animacy. Results show divergent patterns of behavioral and neural responses. Behavioral analyses show that automatic imitation is equivalent when one or both cues to animacy are present but reduces when both are absent. By contrast, right TPJ showed sensitivity to the presence of both animacy cues. Thus, we demonstrate that right TPJ is biologically tuned to control imitative tendencies when the observed agent both looks like and is believed to be human. The results suggest that right TPJ may be involved in a specialized capacity to control automatic imitation of human agents, rather than a universal process of conflict management, which would be more consistent with generalist theories of imitative control. Evidence for specialized neural circuitry that “controls” imitation offers new insight into developmental disorders that involve atypical processing of social information, such as autism spectrum disorders.

INTRODUCTION

Humans automatically imitate other people’s actions, a process that builds rapport, liking, and affiliation between individuals (Chartrand & van Baaren, 2009). Many social factors influence how much we imitate others, such as the appearance, status, and attitudes of interaction partners, as well as our own prior knowledge and beliefs (Heyes, 2011; van Baaren, Janssen, Chartrand, & Dijksterhuis, 2009). However, the neural mechanisms that control the tendency to imitate other people based on social context are not well known. The aim of the current study is to investigate how cues to animacy, specifically physical appearance and belief, influence the neural mechanisms that control automatic imitation.

The neural basis of imitation has been studied extensively, with inferior frontal and parietal cortices implicated in a process that matches observed movements with stored action representations (Iacoboni, 2009; Rizzolatti & Craighero, 2004). The potency of this automatic matching mechanism (i.e., the degree to which observed movements interfere with performed movements) is influenced by perception of human animacy (Gowen & Poliakoff, 2012; Press, 2011). Interference is reduced or absent when an agent has a nonhuman, robotic form (Press, Bird, Flach, & Heyes, 2005; Kilner, Paulignan, & Blakemore, 2003) and also when prior knowledge suggests the agent is not human (Liepelt & Brass, 2010; Longo & Bertenthal, 2009; Tsai & Brass, 2007). These data suggest that both bottom-up (i.e., visual input) and top-down (i.e., knowledge and beliefs) cues to animacy influence the degree to which observed actions are automatically imitated and thus require cognitive control.

Two candidate brain regions for controlling imitation based on perceptions of animacy are anterior medial pFC (mPFC) and right TPJ (rTPJ). These brain regions are consistently engaged in human brain imaging studies of imitation–inhibition (Brass, Ruby, & Spengler, 2009; Brass, Zysset, & von Cramon, 2001). Furthermore, lesions in these areas are associated with disrupted control of imitative tendencies (Spengler, von Cramon, & Brass, 2010; Brass, Derrfuss, Matthes-von Cramon, & von Cramon, 2003), whereas anodal (excitatory) stimulation to rTPJ via transcranial direct current stimulation improves imitation–inhibition performance (Santiesteban, Banissy, Catmur, &

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both anterior mPFC and rTPJ are engaged when reasoning about others’ mental states, such as beliefs, desires, and intentions (Van Overwalle & Baetens, 2009) and altered responses in these regions have been associated with atypical processing of social information in autism spectrum disorders (ASD; Marsh & Hamilton, 2011; Castelli, Frith, Happe, & Frith, 2002). Together, these findings implicate anterior mPFC and rTPJ as key regions for controlling social interactions with other animate, intentional agents (Frith & Frith, 2010; Adolphs, 2009).

In the present fMRI study, we used an imitation–inhibition paradigm devised by Brass, Bekkering, Wohlschlager, and Prinz (2000), which involves performing an action while simultaneously observing a congruent or an incongruent action. RT measurements provide an index of imitative control, as greater cognitive resources are required to inhibit actions that are incongruent than congruent with one’s own intentions (Heyes, 2011; Brass & Heyes, 2005). Characteristics of the observed agent were manipulated to independently vary both bottom–up and top–down cues to human animacy. The bottom–up manipulation varied the physical form of the agent (human vs. robot), whereas the top–down manipulation varied prior instructions about how the observed movements were made (human motion capture vs. a computer animation algorithm). Thus, evaluating the interactions in our factorial design between congruency, form, and belief directly addresses our main research question, which is how cues to animacy influence brain circuits of imitative control (Figure 1). Consistent with previous work (Gowen & Poliakoff, 2012; Press, 2011), we expect animacy cues to increase the potency of automatic imitation. At the neural level, we hypothesize that anterior mPFC and rTPJ will index sensitivity to imitative control based on these cues to human animacy.

METHODS

Participants

Nineteen right-handed students (17 women, 2 men; $M_{\text{age}} = 21.95$ years, $SD = 1.73$ years) of Radboud University Nijmegen participated in this study. All participants were native German speakers and received course credit or € 15 for their participation. Informed consent was obtained in accordance with the local ethics board.

Task and Stimuli

The paradigm was based on prior imitation–inhibition experiments (Brass et al., 2000). The task involved depressing two buttons, one with the index and one with the middle finger of the right hand, and releasing one button depending on the appearance of a numerical cue. When a “1” appeared, participants lifted their index finger, whereas when a “2” appeared, participants lifted their middle finger. The numerical cue was presented between the index and middle fingers of a virtual left hand, which was displayed in a third-person viewpoint. Three aspects in the imitation–inhibition task were manipulated (Figure 1A). First, the congruency between observed and performed finger movements was manipulated. The virtual hand either lifted the same finger as the participant (congruent) or the other finger (incongruent). Second, the agent’s physical form was manipulated by displaying either a human-like or a non-human-like hand. Third, participants’ belief about how the movements were created was manipulated such that they were led to believe they were observing a movement created by real human movements or by a computer program.

Computer animations of index and middle finger lifting movements were created using “Poser 6” animation software (poser.smithmicro.com). The Poser stimuli were constructed using standard key frame animation procedures where the start and the end posture of the movement is defined, and the program interpolates the appropriate movement timing and trajectory to make the movement between the start and end posture as smooth and biological-looking as possible. To manipulate the physical form of the hand, two virtual models were used: a human-like model and a non-human-like model (Figure 1A). The human hand model closely resembled a real human hand, whereas the nonhuman hand model comprised a number of purple blocks that roughly resembled two “fingers” and otherwise bore no resemblance to a human hand. The brightness of the stimuli was matched as closely as possible so that both hands contrasted equally against the background. Only prior to the experiment were participants shown the real Poser animation videos as part of the cover story. During the experiment, participants saw two critical frames that were taken from the full movement animation of the finger lift. The timing of the presentation of the two still images (which, in succession, produced believable apparent motion of a finger lift) were as follows: Frame 1 (initial, resting posture): 1000-msec duration and Frame 2 (final, extended posture): 1500-msec duration. Two versions of each finger lifting movement were created to increase the efficacy of the belief manipulation (see below). One version showed the finger fully stretched in the final position, and one version showed where the finger was not fully stretched in the final position.

To manipulate what participants believed about the observed movements, they were told that the aim of the experiment was to investigate how movements produced through different animation techniques are evaluated and processed in the brain. It was explained that the appraisal of such movements depends critically on the extent to which the observer understands how the animated movements were created. A custom-made 10-min video “documentary” was shown that explained two animation techniques: human motion capture and computer key frame animation. Specifically, participants learned that human motion capture involves recording real human movement via sensors that are attached to the body,
whereas computer key frame animation involves a computer algorithm that fills in intermediate frames of a movement between predefined start and end positions. After the video, participants were asked whether they understood the two techniques and whether they had any questions about the techniques before the experiment started.

Each trial had the following structure. A fixation cross was shown for a jittered duration of 0, 500, 1000, or 1500 msec. A virtual hand was then displayed in a resting position, followed by the final frame from the finger lift video (Figure 1A). Simultaneous to the finger lifting, a number cue was presented between the index and the middle finger of the virtual hand: either “1” or “2.” Participants responded to this cue by lifting either their index or middle finger, which meant they lifted their finger while observing a finger lifting action. Finally, the fixation cross was shown again until the trial duration reached 4.5 sec.

A total of 320 trials were divided into two experimental runs of equal length and evenly filled a 2 (Congruency: congruent, incongruent) × 2 (Form: human, nonhuman) × 2 (Belief: motion capture, computer animation) factorial design. Each run consisted of 20 miniblocks, each comprising eight trials. Each miniblock was preceded with a written cue that informed participants that the movements they were about to see were created using human motion capture or computer animation. Dependent on this message, either the stretched or unstretched finger

Figure 1. (A) Method and design. Each trial had the same structure: A stationary hand was presented first, followed by the appearance of a number cue and a finger lifting movement. Participants responded to the presentation of the number cue with an index finger lift for the number “1” and a middle finger lift for the number “2.” Concurrently, the observed hand either lifted the same (congruent) or a different (incongruent) finger. Two other variables were manipulated: form and belief. The form of the agent could be human or nonhuman. The human hand model closely resembled a real human hand, whereas the nonhuman hand model comprised a number of three-dimensional purple blocks that were arranged to look like two “fingers” and otherwise bore no resemblance to a human hand. In addition, participants were exposed to an elaborate 10-min film that was the cover story to manipulate beliefs about how the videos were made. It was explained the history of motion capture and computer animation techniques and described how the observed movements were produced by recording real human movement or by using a computer algorithm. Together, these conditions filled a 2 (Congruency: congruent, incongruent) × 2 (Form: human, nonhuman) × 2 (Belief: motion capture, computer animation) factorial design. (B) Research questions and analyses. The key contrasts were centered on how the congruency effect (an index of automatic imitation; Heyes, 2011) was modulated by form and belief. (C) Brain ROIs. Red dots show locations of peak responses in previous fMRI imitation-inhibition studies. To produce our ROI search volume, 15-mm radius spheres were centered on these coordinates (see Supplementary Table S1 for more details). The yellow dot is the location of the peak result in the current experiment (see Figure 2 for details). (D) Manipulation check. After the experiment, participants rated on a 10-point scale how realistic (anchors: 1 = very unrealistic, 10 = very realistic) they perceived the finger lift actions to be according to the type of animation technique and hand model. The human hand was rated as more realistic than the non-human-like hand and motion capture actions were rated as more realistic than computer animation actions.
movements were shown during all trials within the block, which means that purported motion capture movements looked slightly different from purported computer animation movements. The assignment of these two movement versions to the belief blocks was counterbalanced across participants so that each movement variation was paired with each instruction equally across the subject sample.

Within each miniblock, two instances of the four possible pairings of congruency and form were shown (congruent action, human hand; incongruent action, human hand; congruent action, human hand; incongruent action, nonhuman hand). Thus, an equal number of congruent and incongruent trials as well as human and nonhuman trials were shown in a random order in each miniblock. Miniblocks were presented pseudorandomly to ensure that a miniblock type was not presented more than three times in succession. After each miniblock, a fixation cross appeared for 10 sec. Because of a technical error, the fixation cross interval was not implemented for six participants.

Before data collection began in the scanner, participants performed 16 practice trials (eight trials from each belief version to the belief blocks was counterbalanced across participants). Behavioral data were analyzed with a 2 (Action Congruency: congruent, incongruent) × 2 (Form: human, nonhuman) × 2 (Belief: motion capture, computer animation) repeated-measures ANOVA. RT was measured as the time between the presentation of the number cue and the relevant button release. Main effects of Congruency, Form, and Belief were calculated. The interaction between Congruency and Form investigated our top–down hypothesis by testing whether a belief that movements were made by human motion compared with a computer algorithm induced a greater congruency effect. Finally, the interaction of all three factors (Congruency, Form, and Belief) tested whether the congruency effect was influenced additionally by the combination of form and belief manipulations. The three-way interaction addresses whether the absence or presence of both human form and a human animacy belief modulate the congruency effect.

**Imaging Data Analysis**

Functional data were analyzed with Statistical Parametric Mapping software (SPM8, Wellcome Department of Cognitive Neurology, London, UK). Functional images were realigned, unwarped, corrected for slice timing, normalized to the Montreal Neurological Institute template with a resolution of $3 \times 3 \times 3$ mm, and spatially smoothed using an 8-mm smoothing kernel. A design matrix was fitted for each participant with nine regressors, one for each cell of the factorial design (eight in total; $2 \times 2 \times 2$, see Figure 1) and one for the message screen prior to each miniblock, which specified the belief manipulation. Stimulus onsets were time-locked to the presentation of the number cue with a duration of zero and convolved with a standard hemodynamic response function.

To examine our two ROIs (anterior mPFC and rTPJ), a mask was generated based on coordinates from prior imitation–inhibition fMRI studies (Figure 1C and Supplementary Table S1). To construct the mask, 15-mm radius spheres were centered on each of these coordinates, and this mask was used to constrain further analyses. Main effects of congruency, form, and belief were evaluated in both directions. We addressed our primary research questions by calculating two-way and three-way interactions between the three factors in our design, as outlined in Figure 1B. The Congruency × Form interaction tested for clusters that showed a greater congruency effect for the human compared with the nonhuman model [(human_incongruent > congruent) > (nonhuman_incongruent > congruent)]. The Congruency × Belief interaction tested for clusters that showed a greater congruency effect for trials within motion capture compared with computer animation blocks [(motioncapture_incongruent > congruent) > (computer_incongruent > congruent)]. The Belief × Form interaction tested for clusters that showed a greater response to motion capture...
than computer animation blocks for a human than nonhuman model [(motioncapture_human > nonhuman) > (computeranimation_human > nonhuman)]. The Congruency × Form × Belief interaction tested for clusters that show a greater congruency effect for human compared with nonhuman form in the motion capture compared with computer animation belief blocks [((motioncapture_human_incongruent > congruent) > (motioncapture_nonhuman_incongruent) > congruent) > (computer_human_incongruent > congruent) > (computer_nonhuman_incongruent) > congruent)]. For completeness, we calculated the inverse of each interaction also and report all findings that pass our thresholding criteria.

In addition to the primary analyses, which are based on the magnitude of BOLD differences between conditions, we also performed an exploratory connectivity-based analysis. Brass and colleagues (2009) have proposed that, rather than acting in isolation, anterior mPFC and rTPJ may act together during the control of imitative responses. This proposal leads to the hypothesis that the responses of anterior mPFC and rTPJ may exhibit functional coupling during the control of imitation. To test the hypothesis that, during the control of imitative responses, the functional relationship between anterior mPFC and rTPJ may be influenced by bottom-up and top-down cues to animacy, we implemented a psychophysiological interaction (PPI) analysis (Friston et al., 1997). PPI can identify brain regions whose activity shows a change in correlation with a seed region as a function of a change in a psychological state. The seed region is the “physiological” element, whereas the psychological state is the “psychological” element of PPI. The PPI analysis was conducted using the SPM generalized PPI toolbox (McLaren, Ries, Xu, & Johnson, 2012). Seed regions were defined based on results that emerged from the primary magnitude-based analysis of BOLD differences between conditions. More specifically, any reliable group-level interaction between congruency and animacy in anterior mPFC or rTPJ would be explored further using PPI. To create seed regions, for each individual participant, we searched within our anterior mPFC–rTPJ mask for clusters showing the interaction between congruency and animacy p < .05 (uncorrected; Spunt & Lieberman, 2012). Volumes were generated using a 6-mm sphere, which was positioned on the peak of each individual’s cluster.

PPI models for each participant included nine PPI regressors, one for each condition from our 2 × 2 × 2 design and one additional regressor for the written cue. To create these regressors, we first specified the time series of the rTPJ seed region as the first eigenvariate. Next, the time series was deconvolved to estimate the underlying neural activity (Gitelman, Penny, Ashburner, & Friston, 2003). This deconvolved time series was multiplied by the predicted, preconvolved time series of each of the nine conditions. This resulted in one PPI for each condition in terms of predicted “neural” activity, which was then convolved with the canonical hemodynamic response function, yielding nine PPI regressors. Each model also included the time series of each condition convolved with the canonical hemodynamic response function and the time series of the seed region as covariates of no interest (McLaren et al., 2012; Spunt & Lieberman, 2012).

Contrast images were taken to the second level for random effects analysis. Correction for multiple comparisons was performed at the cluster level (Friston, Worsley, Frackowiak, Mazziotta, & Evans, 1994), using a voxel-level threshold of p < .005 and 10 voxels and a family-wise error (FWE) cluster-level correction of p < .05. Significant clusters of activity were localized with the SPM Anatomy toolbox (Eickhoff et al., 2005). We only interpret clusters that pass correction for multiple comparisons, but for completeness and to aid future meta-analyses (Lieberman & Cunningham, 2009), clusters that pass the uncorrected threshold are reported in Table 1 as well as within supplementary materials.

RESULTS

Behavioral Findings

A repeated-measures ANOVA with within-subject factors of Congruency (congruent, incongruent), Form (human, nonhuman), and Belief (motion capture, computer animation) was performed on RT data. The log-transformed RTs were used because the distribution of the raw mean average RTs was positively skewed (Tabachnik & Fidell, 1989). Error trials (4.3%) and trials that deviated more than 3 standard deviations from the mean RT within each cell (2.5%) were removed from the analysis. A main effect of Congruency emerged, $F(1, 18) = 124.18, p < .001, \eta^2_p = .87$, with longer RTs during incongruent trials compared with congruent trials (Figure 2A). There was also a main effect of Form, $F(1, 18) = 8.57, p = .009, \eta^2_p = .32$, with longer RTs when the human-like hand was shown compared with the nonhuman hand (Figure 2A). There was no main effect of Belief ($F < 1$).

Because a marginally significant three-way interaction was found between Congruency, Form, and Belief, $F(1, 18) = 4.42, p = .05, \eta^2_p = .20$, we analyzed the Congruency × Form interactions separately within the two belief conditions. This analysis showed that the Congruency × Hand interaction was not significant within motion capture blocks ($F < 1$), but significant within computer animation blocks, $F(1, 18) = 14.86, p = .001, \eta^2_p = .45$. In computer animation blocks, the difference between incongruent and congruent trials was larger when the human hand was shown compared with the nonhuman hand (Figure 2A). When the same data are plotted as a congruency effect (Figure 2B), it is clear that the congruency effect is reduced for the “nonhuman computer animation” trials compared with the other conditions.

Regarding the bottom-up hypothesis that participants will experience more interference when perceiving a human-like hand compared with a nonhuman hand, although the pattern of the means was in line with our
prediction, the results showed that the interaction between Congruency and Hand was not significant ($F < 1$). Regarding our top–down hypothesis, the results showed a significant two-way interaction between Congruency and Belief, $F(1, 18) = 4.72, p = .04, \eta^2_p = .21$. The difference between congruent and incongruent trials was larger during motion capture blocks compared with computer animation blocks (Figure 2A). When including whether participants believed the cover story about the two animation techniques (yes or no) as a between-subject factor in the analysis (12 true believers and 7 possible nonbelievers), we found no significant three-way interaction of this factor with the Congruency × Belief interaction, neither for the whole data nor when analyzing only trials with the nonhuman hand model (both $F < 1$). For this reason, we consider the whole data set in the imaging analyses.

**fMRI Findings**

**Main Effects**

No main effects of Congruency, Form, or Belief emerged that survived correction for multiple comparisons within the two ROIs (anterior mPFC and rTPJ). At the uncorrected threshold ($p < .005, k = 10$), there was a main effect of Congruency (incongruent > congruent) in anterior mPFC and a main effect of Form (human > nonhuman) in anterior mPFC and rTPJ (Table 1 and Supplementary Figure S1). No other main effects were present.

**Interactions**

Neither anterior mPFC nor rTPJ emerged from the two-way interactions measuring Congruency × Form, Congruency ×...
Belief, or Belief × Form at the cluster-corrected threshold. At the uncorrected threshold, a cluster in rTPJ showed a Congruency × Form interaction (Table 1 and Supplementary Figure S1).

One cluster located within rTPJ demonstrated a significant \( (p < .05 \text{ FWE-corrected}) \) three-way interaction measuring Congruency × Form × Belief (Figure 2C). To explore the interaction further based on our a priori predictions that congruency differences would emerge between conditions, we extracted parameter estimates from the cluster peak and compared the difference between incongruent and congruent trials for each condition using four separate paired \( t \) tests. A significant difference was observed between incongruent \((M = 0.27, SE = 0.19)\) and congruent trials \((M = -0.06, SE = 0.19)\) for the “human hand with motion capture” condition, \( t(18) = -2.1, p < .05 \) (Figure 2C). All other \( t \) tests were not significant \((p > .1 \text{ in all cases})\). Thus, rTPJ displayed a greater congruency effect for human than nonhuman trials in motion capture miniblocks compared with computer animation blocks. No response was observed in anterior mPFC at the corrected threshold, but at the uncorrected threshold, one anterior mPFC cluster showed the same three-way interaction (Table 1 and Supplementary Figure S1). Four follow-up paired \( t \) tests comparing congruent and incongruent conditions were all nonsignificant.

Figure 2. Results. (A) Mean RT. A Consistency × Form × Belief repeated-measures ANOVA showed a marginally significant three-way interaction \((p = .05)\). To explore this interaction, we analyzed the Congruency × Form interactions separately within the two belief conditions. This analysis showed that the Congruency × Form interaction was not significant within motion capture blocks \((p > .05)\) but was significant within computer animation blocks \((p < .01)\). In computer animation blocks, the difference between incongruent and congruent trials was larger when the human hand was shown compared with the nonhuman hand. (B) Congruency effect. To visualize the RT data differently, a congruency effect (RTs for incongruent minus congruent trials) was plotted for the form and belief conditions. The congruency effect is reduced for the “nonhuman computer animation” trials compared with the other conditions. (C) Brain imaging data. One cluster located within rTPJ demonstrated a significant \((p < .05, \text{ FWE-corrected for multiple comparisons})\) three-way interaction between Congruency × Form × Belief. To explore the interaction further, we extracted parameter estimates from the cluster peak and compared the difference between incongruent and congruent trials for each condition using four separate paired \( t \) tests. A significant difference was observed between incongruent and congruent trials for the “human hand with motion capture” condition only. All other \( t \) tests were not significant \((p > .05 \text{ in all cases})\). No other main effects or interactions survived correction for multiple comparisons (see Table 1 and Supplementary Figure S1).
To explore the response in rTPJ further, we performed a PPI analysis (McLaren et al., 2012; Friston et al., 1997), which tested the hypothesis that the functional relationship between rTPJ and anterior mPFC is influenced by bottom–up and top–down cues to animacy. This hypothesis was based on the proposal that anterior mPFC and rTPJ work together during the control of imitation (Brass et al., 2009). Our seed selection criteria (see Methods) yielded 17 participants with seed regions. At the initial threshold (\( p < .005, k = 10\) voxels), we found no response within our mask. Reducing the voxel extent to \( k = 5\), we found that rTPJ demonstrated functional coupling with anterior mPFC (Supplementary Table S2 and Supplementary Figure S2).

We also tested one further hypothesis regarding functional coupling during the control of imitation. We hypothesized that, based on animacy cues, rTPJ could influence brain regions associated with the preparation and control of movement, such as frontoparietal mirror system regions and premotor cortex more generally. To evaluate this hypothesis, we extended our search to include the entire brain. This analysis revealed that rTPJ was functionally coupled with a region of dorsomedial pFC, located in Brodmann’s area 6 of premotor cortex, as well as other brain regions that are located outside the motor system (Supplementary Table S2 and Supplementary Figure S2).

Because of the exploratory nature of this PPI analysis, coupled with the fact that our design was not initially chosen (and thus not optimized) for PPI analyses (McLaren et al., 2012), these findings are not discussed further in the main text. However, in Supplementary Figure S2, we offer a preliminary interpretation for the PPI results emerging in brain regions that were hypothesized to show functional coupling with rTPJ. The rationale for including the PPI findings in supplementary information is for these results to provide a starting point for future research investigating the control of imitation using connectivity analyses.

**Manipulation Checks**

A repeated-measures ANOVA was used to compare how realistic participants rated actions according to form (human, nonhuman) and belief (motion capture, computer animation). The results showed that actions of the human hand were rated as more realistic (\( M = 8.5, SE = 0.37\)) than actions of the non-human-like hand (\( M = 5.6, SE = 0.44\)), \( F(1, 18) = 17.02, p = .001, \eta^2_p = .47\). Similarly, motion capture actions were rated as more realistic (\( M = 8.7, SE = 0.34\)) than computer animation actions (\( M = 7.2, SE = 0.48\)), \( F(1, 18) = 21.9, p < .001, \eta^2_p = .54\) (Figure 1D). When the latter was analyzed with the between-subject factor that indicates whether participants believed the cover story (yes or no), no significant interaction with the Belief effect on realism ratings was found, \( F(1, 18) = 2.7, p = .12, \eta^2_p = .13\).

**Psychophysiological Interactions**

To explore the response in rTPJ further, we performed a PPI analysis (McLaren et al., 2012; Friston et al., 1997), which tested the hypothesis that the functional relationship between rTPJ and anterior mPFC is influenced by bottom–up and top–down cues to animacy. This hypothesis was based on the proposal that anterior mPFC and rTPJ work together during the control of imitation (Brass et al., 2009). Our seed selection criteria (see Methods) yielded 17 participants with seed regions. At the initial threshold (\( p < .005, k = 10\) voxels), we found no response within our mask. Reducing the voxel extent to \( k = 5\), we found that rTPJ demonstrated functional coupling with anterior mPFC (Supplementary Table S2 and Supplementary Figure S2).

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**DISCUSSION**

The present findings demonstrate divergence between how the perception of animacy influences neural and behavioral responses during the control of imitation. The data show that rTPJ is sensitive to the presence of both bottom–up and top–down animacy cues, despite no additional interference to RTs. These results suggest that comparable patterns of interference during inhibition of imitation can engage different brain circuits: Only when the agent both looks human and is believed to be human is rTPJ involved. This suggests that rTPJ controls imitation based on who is being observed, both in terms of another’s appearance as well as what is known about him or her. Moreover, our findings suggest that rTPJ is biologically tuned to control automatic imitation of human agents.

**Is rTPJ Specialized for Controlling Interactions with Animate Agents?**

Imitative tendencies are modulated by whom we share social interactions with (van Baaren et al., 2009). For example, we copy others’ actions more if they appear human-like (Gowen & Poliakoff, 2012; Press, 2011) or prosocial (Heyes, 2011), which in turn fosters affiliative relations between individuals (Chartrand & Lakin, 2013). Although inferior frontal and parietal cortices have been consistently implicated in producing imitative responses (Iacoboni, 2009), little is known about the brain circuits that control the tendency to imitate some individuals more than others. Here we show that right rTPJ plays a role in controlling imitation based on who is being observed, both in terms of the agent’s physical appearance as well as what we believe about them.

Brass and colleagues have proposed that rTPJ indexes whether an observed action is attributable to oneself or another agent and that, in conjunction with anterior mPFC, it forms a neural circuit that regulates the distinction between self and other (Brass et al., 2009). In this study, we build upon this cognitive model by demonstrating that rTPJ is sensitive to how human the observed agent is. As such, in controlling imitation, rTPJ distinguishes who or what is “out there” (Frith & Frith, 2010), showing greater sensitivity to agents’ actions that are believed to be made by a human and look human-like. Indeed, if only one cue to human animacy is present, human-like form or a belief that the movement has human origins, the response in rTPJ does not show the same sensitivity to action congruency. This suggests that rTPJ is biologically tuned to resolve conflict between one’s own actions and actions of other humans.

The comparable congruency effect measured in RTs when one or both animacy cues are present is informative to understand the response in rTPJ. First, it replicates prior work, which showed instructions (beliefs) about animacy do not influence imitative tendencies when appearance and kinematics are already human-like (Press,
tasks (Spengler, von Cramon, & Brass, 2009). As such, during imitation mental state reasoning abilities is associated with increases in rTPJ to a greater extent. This suggestion is consistent with prior neuroimaging evidence that rTPJ is involved in social more than nonsocial conflict resolution (Brass, Derrfuss, & von Cramon, 2005). Thus, if an agent is represented as human-like, rTPJ is engaged in conflict resolution between performed and observed actions. In nonsocial cases devoid of a human agent, it is likely that more general conflict resolution mechanisms may operate, such as those based on spatial parameters alone (Brass et al., 2005).

Behavioral studies also support the view that some imitative processes are biologically tuned by showing that imitative compatibility is independent of spatial compatibility (Boyer, Longo, & Bertenthal, 2012; Catmur & Heyes, 2011). Although both compatibility effects could stem from a similar associative learning mechanism (Catmur & Heyes, 2011), rTPJ may only be engaged when others are represented as human (Brass et al., 2005). In cases devoid of a human agent, spatial compatibility processes may be sufficient to resolve conflicting cues. However, the presence of an animate agent may trigger a “social sense” (Kovács, Téglás, & Endress, 2010), which requires additional processes to be engaged during conflict management. For example, rTPJ has been associated with processes that distinguish self from other (Blanke & Metzinger, 2009) as well as with tracking others’ mental states, such as beliefs and desires (Van Overwalle & Baetens, 2009; Saxe, 2006). As a consequence, when interacting with an agent that has human compared with nonhuman attributes, a greater demand may be placed on brain circuits that disambiguate self from other (Brass et al., 2009) or brain circuits that are engaged when spontaneously tracking what others know (Kovács et al., 2010). It could be one or both of these additional processes that engage rTPJ during interactions with human agents, but further research is required to test these proposals.

The emergence of specialized neural circuitry for controlling imitation converges with behavioral evidence from studies of people with ASD, who show intact automatic imitation abilities (Bird, Leighton, Press, & Heyes, 2007) but deficits with controlling imitation based on social contexts (Cook & Bird, 2012). These findings are consistent with accounts of ASD that support a deficit in controlling imitation based on social factors (Southgate & Hamilton, 2008), rather than a deficit in the mirror neuron system (Oberman & Ramachandran, 2007; Williams, White, & Frith, 2009), rather than a primary dysfunction in a system that matches observed and executed actions, such as the mirror neuron system (Iacoboni, 2009). The current findings implicate a role for rTPJ in regulating imitative responses based on the level of human animacy exhibited by agents.

Limitations and Future Directions

Given the subtle complexities of factors that influence imitation during social interaction (Heyes, 2011; van Baaren et al., 2009), rTPJ is unlikely to be the sole “control center” for imitation. Anterior mPFC has been shown to control imitation based on the direction of others’ eye gaze (Wang, Ramsey, & Hamilton, 2011) and a broad set of brain regions have been implicated in the control of imitation based on animacy cues (Gowen & Poliakoff, 2012). Furthermore, in this study, the connectivity analysis provides preliminary evidence for a functional relationship between rTPJ and parts of mPFC during the control of imitation based on animacy cues. Future work should further delineate how imitative tendencies are controlled and how such control influences other brain regions involved in imitation, such as inferior frontal and parietal cortices. Such work could be valuable to disambiguate, or even reconcile, opposing claims regarding neurocognitive dysfunction in ASD, some of which support the view that a mirror neuron dysfunction is present in ASD (Oberman & Ramachandran, 2007; Williams et al., 2001) and others that propose that the control of imitation by social relevance is dysfunctional in ASD (Cook & Bird, 2012).

Conclusion

The present experiment provides novel insight into the brain circuits that regulate who we imitate. We show that in a task that induces a tendency to automatically imitate others’ actions, rTPJ is sensitive to agents that both look human and are believed to be human. This suggests that rTPJ is biologically tuned to provide a signal that is involved with integrating one’s own and others’ actions during social interactions.

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Note

1. Although we performed the analysis on the log transformation of the variables, we report the raw means in Figure 2A.

REFERENCES


