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Abstract

The debilitating effects of social dysfunction in many psychiatric disorders prompt the need for systems-level biomarkers of social abilities that can be applied in clinical populations and longitudinal studies. A promising neuroimaging approach is the animated shapes paradigm based on so-called Frith-Happeé animations (FHAs) which trigger spontaneous mentalizing with minimal cognitive demands. Here, we presented FHAs during functional magnetic resonance imaging to 46 subjects and examined the specificity and sensitivity of the elicited social brain responses. Test–retest reliability was additionally assessed in 28 subjects within a two-week interval. Specific responses to spontaneous mentalizing were observed in key areas of the social brain with high sensitivity and independently from the variant low-level kinematics of the FHAs. Mentalizing-specific responses were well replicable on the group level, suggesting good-to-excellent cross-sectional reliability [intraclass correlation coefficients (ICCs): 0.40–0.99; dice overlap at \( P_{uncorr} < 0.001 \): 0.26–1.0]. Longitudinal reliability on the single-subject level was more heterogeneous [ICCs of 0.40–0.79; dice overlap at \( P_{uncorr} < 0.001 \): 0.05–0.43]. Posterior temporal sulcus activation was most reliable, including a robust differentiation between subjects across sessions (72% of voxels with ICC > 0.40). These findings encourage the use of FHAs in neuroimaging research across developmental stages and psychiatric conditions, including the identification of biomarkers and pharmacological interventions.

Key words: spontaneous mentalizing; fMRI; animated shapes; reliability; biomarker
Introduction

Human mentalizing, also referred to as theory of mind (ToM), includes the recognition of intentions and emotions of a social partner, an ability which has convincingly been related to the interaction of a circumscribed set of neural regions commonly assigned to the ‘social brain’ (e.g. posterior superior temporal sulcus [pSTS], dorsomedial pre-frontal cortex [dmPFC]; Adolphs, 2009). The in-depth characterization of this network with functional magnetic resonance imaging (fMRI) is a major goal of contemporary social and clinical neuroscience. This is reflected by big-data consortia aiming at the identification of the neural mechanisms of social cognition and associated behavioral deficits such as those observed in autism spectrum disorders (ASD; e.g. EU-AIMS consortium; Murphy and Spooren, 2012). A critical prerequisite for suitable neuroimaging biomarkers for developmental and pharmacological research is the assessment of the specificity, reliability and sensitivity of neural responses elicited by mentalizing fMRI paradigms (Loth et al., 2015). However, despite the broad recognition of these requirements (Bennett and Miller, 2010) and the wealth of published fMRI tasks on mentalizing functions (Schurz et al., 2014), no such data is available to date.

Here, we developed an approach to provide systems-level biomarkers of this kind using the so-called Frith-Happé animations (FHAs), a promising candidate tool for the assessment of social brain activations across a broad range of developmental ages and disorder severities. The FHAs are a set of validated video clips depicting simple geometric shapes which interact at different levels of social significance and prompt spontaneous mentalizing functions (Abell et al., 2000). This form of mentalizing occurs automatically, is observable already at an early age, and forms the basis of social functioning in everyday life (Mar and Macrae, 2007; Apperly and Butterfill, 2009). The relevance of spontaneous mentalizing abilities is exemplified by observations that ASD patients can learn to pass other mentalizing tasks based on explicit reasoning, but have difficulties in spontaneous mentalizing clearly map to the social brain (Castelli et al., 2000), questions regarding the specificity and quality criteria of FHA-induced brain responses remain to be addressed.

The current study therefore investigates the specificity, reliability and sensitivity of neural responses to the FHAs. Based on previous evidence (Castelli et al., 2000), we hypothesized that FHA-induced activation of the social brain can be robustly reproduced in a well-powered sample. In order to address the specificity of brain activation in the light of potential confounds, we hypothesized that neural responses to the FHAs in higher order social brain areas are specific to mentalizing processes (as compared to sub-ordinate processes related to animacy and agency perception) and unrelated to the low-level kinematic properties of the animated video clips (Roux et al., 2013). Using a test–retest approach, the reliability of neural responses was assessed in order to inform the use of FHAs in different experimental designs: While cross-sectional reliability is critical for designs involving group comparisons, such as case-control studies (e.g. in imaging genetics or pharmaco-fMRI studies with parallels group designs), longitudinal reliability is required for repeated measures analyses in developmental and cross-over pharmacological designs (Bennett and Miller, 2010). A third type of reliability refers to the aspect of between-subject reliability, which reflects the replicability of between-subject differences and is particularly desirable in endophenotype research (Raemaekers et al., 2007). Since the ‘bottom up’-like nature of spontaneous mentalizing is suggestive of a fairly stable response of the underlying neural circuitry, we expected fair to good (Shrout and Fleiss, 1979) reliability of FHA-induced brain activations. In addition, statistical sensitivity was assessed as a complementary measure and assumed to be high as a result of the block design structure of the task.

Materials and methods

Experimental procedure

Participants. A total of 46 healthy volunteers (mean age: 24.7 ± 5.3 years, 21 females) participated in the study, a subsample of which (n = 28, mean age: 22.9 ± 2.8 years, 14 females) performed fMRI scanning twice in order to assess test–retest reliability estimates. Exclusion criteria included a lifetime history of neurological or psychiatric disorder, current intake of psychoactive substances, significant general medical problems including liver, cardiac, or renal dysfunctions, a history of head trauma, and pregnancy. All individuals provided written informed consent for a study protocol that was approved by the institutional review board of the Medical Faculty Mannheim.

Paradigm. We integrated the FHAs (Abell et al., 2000) in a block-designed fMRI paradigm with a pseudorandomized order of task conditions [random (R), goal-directed (GD), ToM]. All video clips featured a big and a little triangle moving about the screen. In the ToM condition, the triangles’ movement patterns suggested complex intentional interactions challenging the observer’s mentalizing abilities (e.g. one triangle deceiving the other). In the higher order control condition, the triangles interacted purposefully, thereby conveying the perception of agency, but did not challenge mentalizing functions (GD, e.g. one triangle imitating the other). In the second, more basic control condition (not analyzed further here), the triangles moved randomly without interaction (R), thereby only allowing for the perception of animacy. Following the procedures validated by (White et al., 2011), we asked subjects to indicate the subjective social significance level of the depicted interactions (i.e. ToM, GD or R) after each video clip presentation (multiple-choice questions for categorization; MCQ-cat). After the social significance ratings of ToM video clips, we additionally asked the subjects to rate the perceived emotional state (i.e. positive, neutral or negative emotional valence) of each triangle at the end of the video clip presentation (multiple-choice question for the triangle’s feeling; MCQ-feeling). This was done to assess the subjective emotional significance of the scenes and to provide an indicator of whether the expected social cognitive and emotional concept of the displayed interactions had indeed been understood. Further details on the stimuli, trial structure and ratings of the fMRI paradigm are provided in Figure 1. Subjects were thoroughly instructed and trained prior to fMRI scanning using three established practice video clips.

MRI data acquisition. Functional MRI was performed on a 3 T Siemens Trio Scanner (Siemens, Erlangen, Germany) equipped with a 12-channel head coil. Functional data was collected using an echo-planar imaging (EPI) sequence with the following parameters (TE: 30 ms, TR: 2 s, α: 80°, matrix: 64 × 64, FOV: 192 × 192 mm, in-plane resolution: 3 × 3 mm, slice thickness: 4 mm, gap: 1 mm, 28 axial slices, 331 volumes).
Spatial definition of social brain regions. A priori regions of interest (ROI) were defined based on meta-analytical data highlighting their well-established role in human mentalizing. To this end, ROI masks were derived from a recent influential publication on their well-established role in human mentalizing. To this end, ROI masks were derived based on meta-analytical data highlighting the subject’s understanding of the cover story: ‘coaxing’ and ‘surprising’ require both triangles to be rated similarly positively, while ‘mocking’ and ‘seducing’ require the little triangle to be rated more positively than the big triangle. Emotional states were represented by schematic faces with an unhappy, neutral and happy expression, respectively. Bar graphs depict average rating scores (± SE) for each ToM animation, with ‘1’ referring to ‘happy’ and ‘0’ referring to ‘unhappy.’ Panel (C) depicts the temporal structure of a ToM trial. The presentation of the video clips was preceded by a jitter with variable duration (M = 955.67 ms, s.d. = 418.3). Responses were given with the right thumb, using the left, upper and right key of an MRI compatible button box (Current Designs, PA, USA). As soon as responses were given during MCQ ratings, the chosen icon was framed in red for the duration of one additional second, followed by a blank screen for the remainder of the respective MCQ phase. No feedback on response accuracy was given.

Test-retest procedures. Subjects participating in the test-retest study performed all FHAs task procedures twice, including the instructions and practice sessions (mean time interval between the first [T1] and second [T2] session: 15.8 ± 3.5 days). Basic variables such as time of day, hours of sleep, cigarettes smoked and caffeine intake were matched as closely as possible between T1 and T2 in order to control for potential physiological confounds (all P values > 0.05; Supplementary Table S1). In addition, scanner quality assurance (QA) was performed according to an established QA protocol (Plichta et al., 2012). Metrics pertaining to mean signal intensity, spatial and temporal signal-to-noise ratio, percent signal fluctuations and percent signal drift were acquired on every measurement day (n = 37), using the identical EPI sequence parameters as outlined above for a total of 150 volumes. Due to technical issues, QC data was not available for four of the acquired measurements. All QA metrics were stable within the time range of the study (December 2012 to May 2013; Supplementary Table S2).

Data analysis

Functional correlates of spontaneous mentalizing. Image preprocessing followed standard processing routines in SPM8 (http://www.fil.ion.ucl.ac.uk/spm/). Briefly, data was realigned to the first image, slice time corrected, spatially normalized into standard stereotactic space defined by the Montreal Neurological Institute (MNI) template, resampled to 3 mm isotropic voxels, and smoothed with an 8 mm full-width at half-maximum Gaussian Kernel. Individual general linear models (GLMs) included condition-wise regressors, which were calculated by convolving the modelled box-car functions of video clip presentations with the standard canonical hemodynamic response function (HRF) implemented in SPM. Realignment parameters were included as covariates of no interest at the first level. During model estimation, the data was high-pass filtered with a cutoff of 256 s, and an autoregressive model of the first order was applied. In order to separate the functional correlates of mentalizing from those of lower-level processes related to animacy and agency perception, the mentalizing condition (ToM) was contrasted to the higher-level baseline condition (GD). Resulting individual contrast images were subjected to one-sample t-tests for group-level inference. Activations were based on the Anatomical Automatic Labeling Atlas (Tzourio-Mazoyer et al., 2002). See Supplementary Figure S1 for an illustration of the examined ROIs.
reported at a significance level of \( P < 0.05 \), peak-level family-wise error corrected for multiple comparisons across the whole brain.

**Test–retest reliability.** Test–retest reliability was determined for the high-level contrast ToM > GD using the following complementary metrics (described in detail below and illustrated in Figure 2):

1. Dice overlap of thresholded \( t \)-maps (Rombouts et al., 1998; Smith et al., 2005),
2. Spatial reliability, defined as the replicability of the spatial signal intensity distribution derived from contrast maps (Raemaekers et al., 2007), and
3. Single-voxel reliability, defined as the replicability of single-voxel signal intensities across subjects and time points (Caceres et al., 2009).

Cross-sectional and longitudinal reliabilities resulted from the application of the metrics ‘dice overlap’ and ‘spatial reliability’ to group level and single-subject level activation maps, respectively. Between-subject reliability was assessed using the metric ‘single-voxel reliability’ in order to determine how well the rank order of subjects is preserved across sessions. All three metrics were applied to each ROI and to the whole-brain task network (defined by voxels surviving \( P < 0.001 \) at the group level at T1, uncorrected for multiple comparisons across the whole brain; Figure 2D), and calculations were performed as follows:

The intra-class correlation coefficient (ICC) with a consistency criterion (ICC(3,1); Shrout and Fleiss, 1979) was used to assess single-voxel and spatial reliability. Following Fleiss (1986), we denote ICC values \(< 0.40 \) as poor, \( 0.40–0.75 \) as fair to good and \( > 0.75 \) as excellent. Single-voxel reliability was calculated as ICC on signal intensity across subjects and time points for each voxel, and summary values were calculated as percent voxels with ICC \( > 0.4 \) for each ROI. This measure therefore quantifies the proportion of voxels which meet the reliability criterion for each ROI. Spatial reliability was based on contrast maps and was calculated as ICC on signal intensity across voxels and time points. Finally, dice overlap was based on thresholded \( t \)-maps calculated as \( R_{\text{overlap}} = \frac{2 \times A_{\text{overlap}}}{A_1 + A_2} \), which ranges from 0 (no overlap) to 1 (total overlap). The variables \( A_1, A_2 \) and \( A_{\text{overlap}} \) represent the quantity of supra-threshold voxels at T1, T2, and for both sessions, respectively. The overlap was evaluated for two thresholds, \( P < 0.001 \) and \( P < 0.005 \), uncorrected for multiple comparisons across the whole brain, in order to additionally assess the dependence of the overlap measure on the applied threshold. A lack of supra-threshold activation was penalized with assigning a value of zero. In addition, following the assumption that single-subject level reliability should be reflected in higher within- than between-subject overlap (Gorgolewski et al., 2013), session-wise overlap measures were...
computed between each subject and all other subjects and statistically compared to within-subject overlap measures. To this end, the within-subject overlap of each individual was compared to a randomly chosen overlap measure involving the same individual (i.e. between- and within-subject overlap), and an average difference score was calculated. This comparison was repeated 10,000 times. An empirical P value was calculated as the number of occurrences with smaller within-subject overlap (i.e. difference score < 0), divided by the number of repetitions. P values < 0.05 therefore reflect significantly higher within- than between-subject overlap for the respective ROI.

**Power calculations.** In order to assess task sensitivity, i.e. the ability to detect an effect of task in the fMRI signal, power calculations were performed for each ROI separately as well as for all ROIs combined using the FMRIPower toolbox implemented in SPM (fmripay.org; Mumford and Nichols, 2008). Test sample size was set at n = 40, which has been shown to be adequate for the unbiased detection of real effects in fMRI data (Yarkoni, 2009), and Type 1 error rate was set at α = 0.05 for each ROI. The resulting power estimates therefore indicate the probability of detecting an effect at a statistical threshold of P = 0.05 in a future test sample of n = 40 subjects. Consistent with the established standards (Cohen, 1988), power values > 80% were considered as acceptable.

**Effects of low-level kinematics.** To identify and spatially localize the neural effects of low-level kinematics, we used frame-wise kinematic information (Roux et al., 2013) of the triangles’ physical movement properties, namely the ‘instantaneous velocity’ (defined separately for each triangle, \(v_{\text{little}}\) and \(v_{\text{big}}\)) and the ‘relative distance’ (\(D\)) between the triangles. The frame-wise kinematic regressors were i) down-sampled to 0.5 Hz (TR = 2s), ii) \(x\)-transformed (mean 0 and s.d. 1), iii) convolved with the canonical HRF, iv) orthogonalized to the video clip regressors in order to remove the main effect of the animations and v) included as regressors of interest into the first-level (single-subject) models. MCQ ratings were modelled as well to facilitate comparisons against the implicit baseline. Parameter estimation followed the same GLM procedure as described above. Individual beta-images of each kinematic condition were subsequently followed the same GLM procedure as described above. To identify and spatially localize the neural effects of low-level kinematics, we used frame-wise kinematic information (Roux et al., 2013) of the triangles’ physical movement properties, namely the ‘instantaneous velocity’ (defined separately for each triangle, \(v_{\text{little}}\) and \(v_{\text{big}}\)) and the ‘relative distance’ (\(D\)) between the triangles. The frame-wise kinematic regressors were i) down-sampled to 0.5 Hz (TR = 2s), ii) \(x\)-transformed (mean 0 and s.d. 1), iii) convolved with the canonical HRF, iv) orthogonalized to the video clip regressors in order to remove the main effect of the animations and v) included as regressors of interest into the first-level (single-subject) models. MCQ ratings were modelled as well to facilitate comparisons against the implicit baseline. Parameter estimation followed the same GLM procedure as described above. Individual beta-images of each kinematic condition were subsequently subjected to a one-way analysis of variance for second-level statistical inference. Kinematic-sensitive brain regions were identified by testing voxel-wise beta values against the null hypothesis (H0) of no increased response to any of the conditions using the minimum statistic compared to the global null (Nichols et al., 2005). To lower the risk of false negatives (i.e. failure to detect an effect of kinematics in mentalizing-associated brain regions), significance was defined at a very liberal threshold of P < 0.05, uncorrected for multiple comparisons across the whole brain.

**Results**

**Categorization performance**

Mean overall categorization accuracy (calculated as percent correct) was 88.2 ± 7.5% (ToM: 97.8 ± 7.0%; GD: 75.0 ± 14.7%; R: 91.8 ± 14.8%). Emotional ratings of ToM video clips were consistently positive for two animations (coaxing: 0.91 ± 0.28 for the big and 0.81 ± 0.54 for the little triangle; surprising: 0.80 ± 0.59 for the big and 0.93 ± 0.33 for the little triangle), and neutral-to-negative for the remaining two animations (mocking: −0.49 ± 0.59 for the big and −0.10 ± 0.85 for the little triangle; seducing: −0.82 ± 0.48 for the big and 0.11 ± 0.93 for the little triangle; Figure 1).

Given the lower categorization accuracy for GD videos, supplemental analyses were conducted to explore the effects of classification performance on mentalizing-specific brain activation (see Supplementary Material).

**Whole-brain responses to spontaneous mentalizing**

We detected significant activation increases during the mentalizing condition ToM compared to the high-level baseline of GD behavior in areas that were repeatedly highlighted in prior ToM studies (for an overview, see Schurz et al., 2014) and overlapped with the pre-defined social brain masks (see Table 1, Figure 3A). These included pSTS, anterior STS and TP, dMPC, IFG and pre-cuneus. In addition, increased activations were observed in a large cluster extending from the occipital lobe (including anterior and posterior portions of the lateral occipital complex, LO1 and LO2) to the fusiform and inferior temporal gyri.

**Test-retest reliability and power**

While paired t-tests did not reveal any significant changes in group-level brain activations from session 1 to session 2, supplemental analyses were conducted to explore the effects of classification performance on mentalizing-specific brain activation (see Supplementary Material).

**Table 1.** Whole-brain activation during spontaneous mentalizing compared to agency perception (ToM > GD)

| Region                                      | k  | x  | y  | z  | t  | Pcorr |
|---------------------------------------------|----|----|----|----|----|-------|
| Medial temporal pole                        | 1605 | 48 | 11 | −35 | 12.00 | <0.001 |
| Middle temporal gyrus                       | 51  | −58 | 13  | 11.61 | <0.001 |
| Fusiform gyrus                              | 42  | −46 | −17 | 9.58  | <0.001 |
| Superior temporal gyrus [Area PFM ( IPL)]   | 63  | −46 | 22  | 8.34  | <0.001 |
| Middle temporal gyrus                       | 945 | −54 | −52 | 16   | 11.72 | <0.001 |
| Middle occipital gyrus [Area PGP ( IPL)]    | −39 | −79 | 28  | 9.53  | <0.001 |
| Inferior occipital gyrus [hOc4Ip (LO1)]     | 70  | 30  | −94 | 1    | 11.47 | <0.001 |
| Cerebellum [Lobule VIIa crus 2]             | 189 | −21 | −79 | −38  | 10.46 | <0.001 |
| Medial temporal pole                        | 228 | −42 | 14  | 35   | 9.78  | <0.001 |
| Fusiform gyrus                              | 217 | −42 | −49 | 20   | 9.21  | <0.001 |
| Inferior occipital gyrus [hOc4Ip (LO2)]     | −42 | −70 | −8  | 7.00  | <0.001 |
| Cerebellum [Lobule VIIa crus 1]             | 145 | 24  | −79 | −35  | 8.96  | <0.001 |
| Precuneus                                   | 170 | 3   | −55 | 43   | 8.8   | <0.001 |
| Middle occipital gyrus [hOc4la (LO1)]       | 63  | −33 | −94 | −5   | 8.58  | <0.001 |
| IFG (pars orbitalis)                        | 122 | −45 | −29 | −8   | 7.42  | <0.001 |
| IFG                                         | 69  | 57  | 29  | 7    | 6.99  | <0.001 |
| Middle frontal gyrus                        | 51  | −42 | 8   | 52   | 6.88  | <0.001 |
| Superior medial frontal gyrus               | 84  | 3   | 50  | 34   | 6.65  | 0.001 |
| Cerebellum [Lobule IX]                      | 13  | −6  | −55 | −47  | 6.16  | 0.003 |
| Supplementary motor area                    | 15  | 9   | 14  | 67   | 6.01  | 0.004 |
| Precentral gyrus                            | 11  | 45  | 2   | 40   | 5.96  | 0.005 |
| Parahippocampial gyrus                      | 4   | 24  | −4  | −23  | 5.71  | 0.011 |

Note: Cluster extent k is given at Pcorr < 0.05, family wise error corrected for multiple comparisons across the whole brain. Regions were classified according to the Automated Anatomical Labeling Atlas (Tzourio-Mazoyer et al., 2002). If applicable, anatomical labels were added in square brackets based on Anatomical Probability Maps (Anatomy toolbox; Eickhoff et al., 2006). X-, y-, and z-coordinates MNI and statistical information refer to the peak voxel(s) in the corresponding cluster. P values are adjusted for family wise error correction for multiple comparisons across the whole brain.
reliabilities were heterogeneous across the different ROIs and metrics (see Table 2, Figure 3 B–D). Highest values were obtained for dice overlap and spatial reliability on the group level, indicating high cross-sectional reliability across ROIs. On the single-subject level, no ROI fell below the lower bound (0.40) of the defined range of acceptable ICC values for spatial reliability (calculated as the median of individual ICCs), suggesting that spatial activation patterns are sufficiently reliable in a longitudinal setting. Dice overlap for single-subject maps was more heterogeneous (maximum of 0.43 and 0.53 for the right pSTS at a threshold defined at $P < 0.001$ and $P < 0.005$, respectively). Despite this heterogeneity, the additional criterion of greater within- than between-subject overlap was met by most ROIs. Between-subject reliability as assessed as single-voxel reliability was also heterogeneous across ROIs (overall mean of 34% reliable voxels). A reliable portion of at least 10% of voxels was found for each ROI except for the precuneus. Larger portions were found in the pSTS, right TP, right IFG and left aMTG (47%–72%). Of note, the clustering of reliable voxels across the whole brain did not stringently overlap with the activation (Figure 3 A and B) or deactivation peaks (data not shown). Similar observations have been reported in previous reliability studies (Caceres et al., 2009; Plichta et al., 2012). Power calculations revealed that all ROIs performed well above the commonly adopted 80% threshold (Cohen, 1988).

**Differential responses to low-level kinematics**

Effects of low-level kinematic stimulus properties were mainly observed in areas of the dorsal visual pathway and the downstream oculomotor network (Goodale and Milner, 1992; Petit and Haxby, 1999), including the motion-sensitive extrastriate areas V3A and V5, superior parietal lobe (SPL) and frontal eye fields (FEF; at the caudal end of the superior frontal sulcus; Table 3 and Figure 4A). Visual comparison of whole-brain effects of low-level kinematics to both pre-defined ROIs (Figure 4B) and whole-brain observed (Figure 4C) mentalizing effects suggested largely non-overlapping activation patterns (no overlap for all social brain ROIs, below 6% for the T1 network at $P < 0.05$, uncorrected for multiple comparisons across the whole brain, see panels B and C in Figure 4).

**Discussion**

The human social brain is the focus of many neuroimaging studies dedicated to basic neuroscience and clinical research. Given the high clinical relevance of social dysfunction and the advance of large multicenter fMRI studies, empirical knowledge on the quality criteria of social tasks becomes increasingly important. To this end, the current study investigated the specificity, reliability and sensitivity of brain responses to the FHAs, a well-established set of experimental stimuli arising from cognitive psychology (Abell et al., 2000). Among others, we demonstrate a strong differential engagement of key structures of the human social brain to stimuli engaging mentalizing processes relative to the high-level control stimuli challenging simple intention detection. We hope that our findings, discussed in more detail below, will be useful to guide the application of FHAs in ongoing large-scale studies aiming at identifying the neural mechanisms of normal and altered social cognition.

As our first main finding, we observed that the ToM video clips provoked a strong activation increase in brain regions that have been previously implicated in various aspects of social cognition. While the strongest responses mapped to...
|                  | Session-wise activation: peak contrast value | Group map: cross-sectional reliability | Single subject map: longitudinal reliability | Single voxel: between-subject reliability | Power |
|------------------|---------------------------------------------|---------------------------------------|---------------------------------------------|------------------------------------------|-------|
|                  | S1 (s.d.) | S2 (s.d.) | P value\(^a\) | spatial ICC ± 95% CI | overlap at P < 0.001 | overlap P < 0.005 | median spatial ICC (IQR) | mean within overlap at P < 0.001 | P value\(^b\): within vs. between overlap at P < 0.005 (s.d.) | P value\(^b\): within vs. between overlap at P < 0.005 | ICC > 0.40 |
| T1 network\(^c\) | 1.59 (0.56) | 1.59 (0.47) | 0.988 | 0.9 (0.90 0.91) | 0.89 0.93 | 0.54 (0.41 0.65) | 0.34 (0.22) | <0.001 | 0.42 (0.22) | <0.001 | 36% 99.97 |
| pSTS R           | 1.15 (0.59) | 1.15 (0.48) | 0.995 | 0.98 (0.97 0.98) | 1 1 | 0.69 (0.49 0.83) | 0.43 (0.32) | 0.008 | 0.53 (0.31) | 0.016 | 72% 100 |
| pSTS L           | 0.98 (0.39) | 0.85 (0.32) | 0.085 | 0.99 (0.98 0.99) | 0.99 1 | 0.79 (0.65 0.82) | 0.43 (0.33) | <0.001 | 0.52 (0.34) | 0.002 | 43% 100 |
| dmPFC            | 0.94 (0.61) | 0.84 (0.41) | 0.517 | 0.95 (0.94 0.96) | 0.84 0.9 | 0.6 (0.40 0.74) | 0.11 (0.17) | 0.024 | 0.19 (0.21) | 0.007 | 11% 83.79 |
| preC             | 0.74 (0.35) | 0.74 (0.29) | 0.933 | 0.97 (0.96 0.97) | 0.96 0.97 | 0.56 (0.24 0.79) | 0.2 (0.31) | <0.001 | 0.25 (0.33) | 0.002 | 6% 98.53 |
| IFG R            | 0.75 (0.39) | 0.87 (0.32) | 0.163 | 0.96 (0.95 0.98) | 0.85 0.92 | 0.66 (0.41 0.87) | 0.25 (0.3) | 0.016 | 0.36 (0.3) | 0.01 | 48% 97.51 |
| IFG L            | 0.77 (0.44) | 0.91 (0.45) | 0.282 | 0.89 (0.82 0.93) | 0.92 0.93 | 0.61 (0.33 0.77) | 0.16 (0.29) | 0.043 | 0.25 (0.32) | 0.097 | 14% 99.97 |
| aMTG R           | 0.56 (0.26) | 0.55 (0.27) | 0.942 | 0.85 (0.78 0.90) | 0.89 0.94 | 0.59 (0.45 0.73) | 0.19 (0.28) | 0.003 | 0.24 (0.33) | 0.013 | 18% 99.95 |
| aMTG L           | 0.61 (0.22) | 0.36 (0.27) | 0.351 | 0.73 (0.60 0.82) | 0.7 0.8 | 0.45 (0.16 0.65) | 0.06 (0.16) | 0.048 | 0.14 (0.23) | 0.014 | 47% 99.42 |
| TP R             | 0.77 (0.25) | 0.74 (0.31) | 0.608 | 0.64 (0.53 0.72) | 0.64 0.66 | 0.49 (0.18 0.63) | 0.31 (0.3) | <0.001 | 0.36 (0.32) | <0.001 | 49% 100 |
| TP L             | 0.8 (0.37) | 0.65 (0.42) | 0.147 | 0.4 (0.21 0.56) | 0.26 0.43 | 0.4 (0.10 0.59) | 0.05 (0.14) | 0.161 | 0.11 (0.19) | 0.024 | 33% 100 |

\(^a\)P value as assessed by paired t-test on session-wise peak values.

\(^b\)P value calculated as empirical P value from 10 000 repetitions (see Materials and methods).

\(^c\)T1 network consisted of 8147 voxels.

S1: session 1, S2: session 2, ICC: intraclass correlation coefficient, R overlap: spatial overlap measure, CI: confidence interval, s.d.: standard deviation, IQR: interquartile range (Q1 Q3), ES: effect size (calculated as Cohen’s d), pSTS: posterior superior temporal sulcus, dmPFC: medial prefrontal cortex, precuneus, IFG: inferior frontal gyrus, aMTG, anterior middle temporal gyrus, TP: temporal pole, L: left, R: right.
occipito-temporal areas involved in social perception (Allison et al., 2000; Deen et al., 2015), the activations of the inferior parietal lobule and IFG are consistent with the self-referential representation of the observed actions (e.g. mirroring; Iacoboni and Dapretto, 2006). In addition, the ToM animations recruited structures known for subserving cognitive functions that allow for the higher-order representation of complex social scenes, such as perspective taking (precuneus; Cavanna and Trimble, 2006), abstract reasoning (dmPFC; Bzdok et al., 2013), and social knowledge retrieval (TP; Olson et al., 2013). The observed breadth of activations is hereby in line with current network accounts on social cognition, which propose that specialized but highly interrelated neural circuits operate on an implicit-to-explicit continuum (Yang et al., 2015). Applied to our findings, this framework would suggest that the strong engagement of social-perceptive areas reflects the implicit bottom-up demands

Table 3. Whole-brain activation to low-level kinematics

| Region                                      | k  | x     | y     | z     | t    | P corr |
|---------------------------------------------|----|-------|-------|-------|------|--------|
| Middle occipital gyrus [hOc4lp (LO2), hOc5 (V5/MT)] | 575| -45   | -79   | 13    | 6.19 | <0.001 |
| Superior occipital gyrus [hOc4d (V3A)]      | 1376| 15    | -61   | 67    | 5.32 | <0.001 |
| Superior parietal lobule [Area 7A (SPL)]    | -18| -55   | 67    | 4.11  |      |        |
| Middle temporal gyrus [hOc4lp (LO2), hOc5 (V5/MT)] | 262| 45    | -73   | 7     | 5.08 | <0.001 |
| Lingual gyrus [hOc1 (V1), hOc2 (V2)]        | 68 | 9     | -91   | -8    | 3.70 | <0.001 |
| Precentral gyrus (FEF, BA 6)                | 327| 24    | -13   | 58    | 3.61 | <0.001 |
| Superior frontal gyrus (FEF, BA 6)          | 370| -21   | -10   | 55    | 3.22 | <0.001 |
| Cerebellum [Lobule VIIIa]                   | 19 | -30   | -40   | -47   | 1.45 | 0.983  |
| Cerebellum [Lobule VIIIa]                   | 16 | 33    | -40   | -44   | 1.02 | 1.000  |
| Rectal gyrus                                 | 4  | 9     | 47    | -20   | 0.97 | 1.000  |
| Medial temporal pole                         | 1  | 39    | 20    | -38   | 0.35 | 1.000  |

Note: Cluster extent k is given at $P < 0.05$ (uncorrected for multiple comparisons across the whole brain). BA, Brodmann Area, FEF, frontal eye field. Regions were classified according to the Automated Anatomical Labeling Atlas (Tzourio-Mazoyer et al., 2002). If applicable, anatomical labels were added in square brackets based on Anatomical Probability Maps (Anatomy toolbox; Eickhoff et al., 2006). X-, y-, and z-coordinates MNI and statistical information refer to the peak voxel(s) in the corresponding cluster. P values are adjusted for family wise error correction for multiple comparisons across the whole brain. Note that all clusters were identified at a liberal voxel-wise threshold of of $P < 0.05$, uncorrected for multiple comparisons across the whole brain, in order to maximize sensitivity for potential confounding effects.

Fig. 4. Whole-brain effects of low-level kinematic stimulus properties. (A) Sections displaying activated clusters. (B) Sections showing outlines of the pre-defined mentalizing network (red), overlaid on whole-brain effects of low-level kinematics. (C) Outlines of the whole-brain effects of mentalizing (ToM > GD). All statistical maps were thresholded at $P < 0.05$, uncorrected for multiple comparisons across the whole brain to minimize false negatives.

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of spontaneous mentalizing. Conversely, the engagement of higher-order areas likely reflects the more explicit demands of the stimuli, for example, those related to the evaluation and rating of the ToM video clips (MCQ-cat, MCQ-feelings). We thus conclude from these data that, albeit their minimal task demands, the FHAs are suitable for challenging both bottom-up sensory and higher-order representational areas and processes in the social brain. Moreover, in contrast to purely passive tasks, the implemented non-verbal ratings provide a form of behavioral control during data acquisition (e.g. to secure that the subjects indeed performed the task and understood the presented social scenarios). For instance, the observed category ratings imply that the subjects recognized the different levels of social significance of the presented interaction. Of note, our supplemental analyses suggest that misclassifications can meaningfully explain activation within the mentalizing network (see Supplementary Material). In addition, the valence ratings suggest that the emotional significance (or leitmotif) of the cover stories was generally understood, with a certain variability across subjects that may point to (potentially clinically and sub-clinically meaningful) differences in the acquisition of the social emotional concepts of the animations. We expect these behavioral measures to be very useful in future studies since they may, for example, guide the identification of the neural correlates of individual differences in subjective stimulus evaluation and allow for the comparison of performance-adjusted groups.

As a second study goal, we examined the functional correlates of the low-level kinematics of the FHAs. Prior work identified kinematic differences in form of higher-immobilization rate and lower relative distance of the animated shapes in the ToM conditions, which related to differences in the eye-movement towards the FHAs conditions (Roux et al., 2013). With respect to neural network activation, our data suggests that the predominant effect of the stimulus kinematics is rather localized and is confined to the visuo-oculomotor circuitry (Petit and Haxby, 1999). Notably, differences in the kinematic profile are an inherent determinant of the social significance of the FHAs (and plausibly also of natural social interactions) and are therefore not fully separable from the video clip conditions (Scholl and Tremoulet, 2000; Roux et al., 2013). However, our results suggest that their relevance to higher order social brain responses to the task are rather low. Importantly, the kinematic-specific responses did not even overlap with the social condition-specific effects within visual areas. The latter were observed in the shape-selective area LO1 as well as in the mid-fusiform gyrus, both representing higher order visual areas involved in the detection of agency and social meaning (Shultz and McCarthy, 2014; Malkovik et al., 2015).

Besides the specificity of the neural responses, we studied the task’s capability to robustly activate the targeted system by means of test-retest reliability assessments along with power calculations of the elicited neural signals. In principle, a reliable task allows the attribution of detected signals to factors other than unstable task effects. The reproducibility of fMRI data can be compromised by multiple factors, such as changes in acquisition parameters (e.g. scanner hardware, field strength, image signal-to-noise ratio) and subject-related factors (e.g. habituation, motion, cognitive strategies; Raemaekers et al., 2007; Caceres et al., 2009; Bennett and Miller, 2010; Gorgolewski et al., 2013). Prior studies have assessed the reliability of brain responses in various domains, including emotion and motivation (e.g. Johnstone et al., 2005; Plichta et al., 2012, 2014; Lipp et al., 2014), executive (e.g. Caceres et al., 2009; Plichta et al., 2012) and sensorimotor functions (e.g. Zandbelt et al., 2008; Caceres et al., 2009), but not for mentalizing tasks targeting the social brain. Notably, unlike many prior reliability studies, we specifically focused on the robustness of the high-level experimental contrast, controlled for a range of basic physiological confounds, and aimed at maximizing the generalizability of results by constraining our analysis to a set of meta-analytically derived ROIs.

We applied two different metrics, dice overlap and spatial reliability, to the group- and single-subject data in order to quantify aspects of cross-sectional and longitudinal reliability, respectively. These analyses revealed good-to-excellent reliability of group-level responses to FHAs across ROIs and metrics, which is in line with the observation of a generally high reproducibility of other robust fMRI paradigms studied cross-sectionally (Plichta et al., 2012). Longitudinal reliability was lower, in particular for the spatial overlap measure, although the criterion of greater within- than between-overlap of supra-threshold activation (Gorgolewski et al., 2013) within predefined ROIs was met by most areas of the social brain. Spatial reliability was fair-to-good for all ROIs, and even excellent for the right pSTS. These results are in line with previous reports of lower reliability of fMRI results obtained on the single-subject compared to the group level (Raemaekers et al., 2007; Plichta et al., 2012; Lipp et al., 2014).

Across ROIs, the overall reliability pattern suggests that activity was less reproducible in areas linked to higher-order cognitive processes, such as meta-cognition and self-reflection (e.g. dmPFC, precuneus; Cavanna and Trimble, 2006; Bzdok et al., 2013). A plausible reason for this observation is the fact that the triangle animations allow for a certain degree of ‘cognitive freedom’ (Gorgolewski et al., 2013) since they allow to be interpreted differently by different subjects and might therefore be associated with a greater variability of neural responses. The reliability of the pSTS, in contrast, was excellent across levels (i.e. single-subject and group level) and metrics (i.e. spatial reliability and dice overlap), which is in line with this region’s involvement in more implicit, bottom-up and thus possibly more invariant processes linked to social perception (Deen et al., 2015). The repeated exposure of FHAs stimuli within a two-week interval might therefore not influence bottom-up processes responsible for triggering spontaneous mentalizing, but may have an impact on how the videos are interpreted upon second presentation (e.g. no need for a de-novo reconstruction of the presented social scenario). In more general terms, our finding of different reliabilities across ROIs likely reflects the difference between more constrained bottom-up and more flexible top-down processing of social information.

The right pSTS also yielded the best differentiation between subjects, with more than 70% of voxels displaying at least fair-to-good between-subject reliability (i.e. single-voxel reliability with ICC(3,1) > 0.40). Since greater heterogeneity between subjects leads to higher ICC values, our results suggest that pSTS activity, besides being highly consistent across sessions, is sufficiently heterogeneous across subjects and is thus particularly suitable for endophenotype research (Raemaekers et al., 2007).

Besides test-retest reliability, statistical sensitivity was assessed using power calculations for a future test sample of 40 subjects. Here, all key regions of the social brain showed a very good probability to detect a functional effect while keeping the type I error rate below 5%. The good statistical sensitivity may be a consequence of the bottom-up driven processes of FHAs, but also the robust block structure of the task and the higher signal-to-noise ratios which result from the chosen voxel size (Bennett and Miller, 2010) and smoothing kernel (Caceres et al., 2009).
Following the systems biology approach, the aim of task-based fMRI in biomarker research is to access specific functional networks with sufficient reliability and sensitivity in order to identify markers of normal or pathogenic processes or of treatment response (Biomarkers Definitions Working Group, 2001). While the obtained fMRI phenotypes may, to some extent, be specific for the employed task (e.g. spontaneous mentalizing vs. false-belief reasoning) and associated behavior, they reflect the interplay of large-scale and oftentimes overlapping neural circuitries. Aberrant activation patterns in patient groups can therefore point to a disruption of information processing within or between these networks. According to this rationale, we believe that our neuroimaging findings support the following conclusions and recommendations for biomarker research on human mentalizing processes:

1. The FHAs lead to a relatively specific activation of social brain areas including networks involved in social perception, action observation and ToM. The high-level contrast of the task allows for the separation of effects of mentalizing from those related to simple intention attribution. Although the observed condition-dependent differences in activation appear to be mainly quantitative, qualitative differences might emerge in clinical populations such as patients with ASD (Castelli et al., 2002).

2. Within our fMRI paradigm and the associated data acquisition and analysis procedures, the FHAs allow for a reliable characterization of social brain activations in cross-sectional and—to a certain extent—also longitudinal study designs. These stimuli therefore represent a promising means to identify and stratify imaging biomarkers and validate pharmacological interventions (Loth et al., 2015), although their use in clinical populations additionally requires to demonstrate reliably altered brain responses in these samples.

3. The pSTS additionally demonstrated high between-subject reliability and is therefore particularly suited for tracing neural correlates of inter-individual differences, such as learning-related differences or potential genetically influenced neural traits (Raemaekers et al., 2007). For instance, the repeated observation of pSTS dysfunctions in highly heritable disorders such as ASD (e.g. Zilbovicius et al., 2006) supports the proposed value of this region for genetic research.

Finally, the specific and reliable brain responses to FHAs encourage the evaluation and use of this task for the characterization of mentalizing-specific functional connectivity differences in clinical research. We expect this line of research to offer insights into preserved and altered human social brain network dynamics in health and disorders with prominent social dys-function such as schizophrenia and autism.

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Supplementary data

Supplementary data are available at SCAN online.

Conflict of interest. None declared.

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