# Supplementary material

for

Domain-specific and domain-general neural network engagement during human-robot interactions

by

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Table S1. Coordinates for the Theory-of-Mind network									
Region	MNI Coordinates								
	x	у	Z.						
Dorsomedial prefrontal cortex	-6	54	36						
Middle medial prefrontal cortex	-4	58	16						
Ventromedial prefrontal cortex	-4	56	-16						
Precuneus	0	-54	34						
Left temporoparietal junction	-48	-62	30						
Right temporoparietal junction	48	-60	30						

Coordinates are based on Richardson et al. (2018).

Region	MNI Coordinates					
	x	у	Z.			
Left occipital face area	-40	-76	-18			
Right occipital face area	44	-76	-12			
Left fusiform face area	-40	-52	-18			
Right fusiform face area	38	-42	-22			
Left posterior superior temporal sulcus	-54	-36	6			
Right posterior superior temporal sulcus	48	-38	4			
Left extrastratiate body area	-48	-74	10			
Right extrastratiate body area	50	-70	10			

## Table S2. Coordinates for the person perception network

Coordinates are based on Julian et al. (2012).

Region	MNI Coordinates					
	x	у	Z			
Left fusiform gyrus	-26	-60	-10			
Right fusiform gyrus	30	-50	-8			
Left lateral occipital complex	-46	-72	-4			
Right lateral occipital complex	46	-70	-4			
Left superior parietal lobule	-24	-56	60			
Right superior parietal lobule	24	-52	64			
Left middle occipital gyrus	-30	-84	12			
Right middle occipital gyrus	36	-84	12			

#### Table S3. Coordinates for the object-selective regions

Bilateral lateral occipital complex and superior parietal lobule coordinates are based on Julian et al. (2012), while the bilateral fusiform gyrus, superior parietal lobule, and middle occipital gyrus coordinates are based on Henschel, Hortensius and Cross (2020) using independent coordinates from Dubey et al. (2020).

Region	MNI Coo	rdinates	
	x	у	Z
Left orbital portion of the inferior frontal gyrus	-42	28	-4
Left inferior frontal gyrus	-48	18	20
Left middle frontal gyrus	-44	-6	58
Left anterior temporal lobe	-52	-8	-14
Left posterior temporal lobe	-48	-38	2
Left angular gyrus	-48	-62	20
Right orbital portion of the inferior frontal gyrus	46	30	4
Right inferior frontal gyrus	42	18	22
Right middle frontal gyrus	50	-4	52
Right anterior temporal lobe	48	-4	-18
Right posterior temporal lobe	44	-32	0
Right angular gyrus	46	-62	20

### Table S4 Coordinates for the language network

Coordinates of the regions are based on the peak voxels of the group maps masked with the network parcels using data from Diachek and colleagues (2020; Experiment 1, n = 383, https://osf.io/pdtk9/).

Region	MNI Coordinates				
	x	у	Z.		
Left posterior parietal lobe	-16	-66	32		
Left middle parietal lobe	-44	-54	42		
Left anterior parietal lobe	-52	-44	48		
Left superior frontal gryus	-18	16	62		
Left precentral gyrus	-40	18	42		
Left opercular portion of the inferior frontal gyrus	-34	18	34		
Left middle frontal gyrus	-40	42	14		
Left orbital middle frontal gyrus	-40	46	10		
Left insula	-36	14	-6		
Left medial frontal gyrus	-2	26	40		
Right posterior parietal lobe	18	-62	32		
Right middle parietal lobe	54	-52	40		
Right anterior parietal lobe	54	-46	40		
Right superior frontal gryus	22	18	60		
Right precentral gyrus	40	18	50		
Right opercular portion of the inferior frontal gyrus	38	18	40		
Right middle frontal gyrus	38	22	48		
Right orbital middle frontal gyrus	42	48	-2		
Right insula	30	18	-12		
Right medial frontal gyrus	2	32	34		

## Table S5 Coordinates for the multiple-demand network

Coordinates of the regions are based on the peak voxels of the group maps masked with the network parcels using data from Diachek and colleagues (2020; Experiment 1, n = 383, https://osf.io/pdtk9/).

Network	Main effect of dynamics			Main effect of type	of interaction		Interaction between dynamics and type of interaction		
	F	р	$\eta^2_G$	F	р	$\eta^2_G$	F	р	$\eta^2_G$
Theory-of-Mind	101.48	1.703e-09	0.37	3.25	.086	0.03	2.28	.146	0.01
Person perception	96.60	2.621e-09	0.55	7.47	.013	0.04	24.75	6.363e-05	0.09
Object-specific	1.45	.242	0.01	6.69	.017*	0.07	1.60	.219	5.07e-03
Language	29.18	2.338e-05	0.23	2.95	.1	0.01	51.50	4.487e-07	0.10
Multiple-demand	18.47	.0003	0.09	0.38	.545	4.05e-03	0.33	.572	1.17e-03

Table S6. Comparison of activity within each network when listening and speaking during interactions with a human and robotic agent

Beta estimates were extracted for each event for each ROI and averaged across network and entered in a 2 (interaction: HHI, HRI) by 2 (dynamics: listening, speaking) repeated-measures ANOVA for each network separately. Significant effects are in bold, \* Did not survive Bonferroni correction (p < .05 / 5);  $\eta^2_G$ : generalised eta-squared.

Network	Main effect of	n effect of dynamics		type of interaction	Interaction bet	Interaction between dynamics and type of interaction		
	Estimates	CI (95%)	Estimates	CI (95%)	Estimates	CI (95%)		
Theory-of-Mind	1.99	1.34 - 2.65	0.14	-0.51 - 0.80	0.78	-0.08 - 1.67		
Person perception	2.60	2.13 - 3.05	0.08	-0.38 - 0.54	-0.96	-1.600.32		
Object-specific	0.37	-0.11 - 0.87	-0.43	-0.92 - 0.06	-0.27	-0.94 - 0.40		
Language	1.66	1.25 - 2.07	0.34	-0.06 - 0.75	-1.14	-1.710.58		
Multiple-demand	0.44	0.08 - 0.80	0.05	-0.32 - 0.40	0.12	-0.37 - 0.61		

The following linear Bayesian regression model was specified: value ~ dynamics.d \* interaction.d + (1|sub), with value representing the beta estimates extracted for each event for each ROI averaged across network, dynamics and type of interaction as fixed effects and a random intercept for participants (sub). Deviation coding style was used with 0.5: HHI/listening, and -0.5: HRI/speaking. Negative values for the main effects indicate a bias towards HRI or speaking, while positive values indicate a bias towards HHI or listening. CI (95%): 95% credibility interval

Network-network	With g	lobal sig	nal regression	Without g	global signa	al regression	Bayesian estimation		
connectivity									
	$M_{diff}$	t	95% CI	$M_{diff}$	t	95% CI	Estimates	CI (95%)	
tom-tom	-0.00	-0.28	-0.02, 0.02	-0.02	-1.68	-0.04, 0.00	-0.01	-0.07 - 0.04	
ppn–ppn	0.01	0.83	-0.01, 0.02	-0.00	-0.14	-0.02, 0.0]	-0.01	-0.06 - 0.03	
object-object	0.00	0.62	-0.01, 0.02	-0.00	-0.15	-0.02, 0.02	-0.02	-0.06 - 0.01	
language-language	-0.00	-0.80	-0.01, 0.00	0.00060	0.09	-0.01, 0.01	-0.01	-0.04 - 0.03	
demand-demand	0.01	2.02	-0.00032, 0.02	0.01	1.25	-0.01, 0.02	-0.02	-0.05 - 0.01	
tom-ppn	0.01	1.32	-0.01, 0.02	-0.00	-0.15	-0.02, 0.01	-0.03	-0.07 - 0.01	
tom-object	0.01	1.14	-0.01, 0.02	-0.00	-0.57	-0.02, 0.01	-0.02	-0.05 - 0.02	
tom-language	0.01	2.65*	0.00, 0.02	0.02	2.19*	0.00092, 0.04	-0.02	-0.06 - 0.01	
tom-demand	0.00	1.03	-0.00, 0.01	0.00	0.18	-0.01, 0.02	-0.02	-0.05 - 0.01	
ppn-object	0.01	1.73	-0.00, 0.03	0.01	0.63	-0.01, 0.02	-0.03	-0.06 - 0.01	
ppn-language	0.00	1.10	-0.00, 0.01	0.01	1.03	-0.01, 0.02	-0.02	-0.06 - 0.01	
ppn-demand	0.01	2.86*	0.00, 0.02	0.01	1.23	-0.01, 0.03	-0.02	-0.05 - 0.00	
object-language	0.00	0.32	-0.01, 0.01	0.00	0.32	-0.01, 0.01	-0.01	-0.04 - 0.03	
object-demand	0.01	3.16*	0.00, 0.02	0.01	1.35	-0.01, 0.03	-0.03	-0.06 - 0.00	
language-demand	0.01	2.18*	0.00053, 0.02	0.01	0.78	-0.01, 0.02	-0.02	-0.06 - 0.01	

Table S8.	Comparison	of functional	connectivity	v between inter	ractions with	a human and	robotic agent
			•/				

Paired-sample t-test was used to test for differences in functional correlation (Fisher z-transformation) between human-human and human-robot interactions for each within- and between-network combination. To test for robustness, these analyses were rerun with nuisance regression without global signal regression (but including nuisance signals from white matter, cerebrospinal fluid and global signal sources (Liu et al., 2017). For Bayesian re-analysis, the following model was specified: corz ~ run \* interaction + (run | sub), with corz representing the Fisher z-transformed Pearson's correlation coefficients between the time courses for all possible combinations of ROIs within or between the network(s) with type of interaction (HHI, HRI) and run (1-4) as fixed effects and a random intercept and slope for participants (run | sub). Estimated posterior regression coefficient for the comparison between human-human and human-robot interaction are reported. \* Did not survive Bonferroni correction (p < .05 / 15);  $M_{diff}$ : mean difference; 95% CI: 95% confidence interval for  $M_{diff}$ ; tom: Theory-of-Mind; ppn: person perception; object: object-specific; demand: multiple-demand

Network	Main effect of run			Main effect of t	Main effect of type of interaction			Interaction between run and type of interaction HRI		
	F	p	$\eta^2_G$	F	р	$\eta^2_G$	F	р	$\eta^2_G$	
Theory-of-Mind – Theory-of-Mind	1.48	.24	0.03	4.223044e-06	.998	1.001184e-08	0.81	.48	0.005	
Person perception – Person perception	0.37	.71	0.01	1.14	.30	0.002	0.09	.94	0.0006	
Object-specific – Object-specific	3.12	.04*	0.05	0.18	.68	0.0005	1.28	.29	0.008	
Language – Language	0.44	.70	0.01	0.45	.51	0.0005	0.74	.51	0.003	
Multiple-demand – Multiple-demand	0.89	.44	0.02	4.39	.05	0.008	0.38	.75	0.002	
Theory-of-Mind – Person perception	0.44	.67	0.01	2.57	.12	0.004	0.48	.68	0.002	
Theory-of-Mind – Object-specific	1.24	.30	0.02	1.43	.24	0.003	0.54	.65	0.003	
Theory-of-Mind – Language	1.13	.34	0.02	6.92	.02*	0.008	0.32	.79	0.002	
Theory-of-Mind – Multiple-demand	0.80	.49	0.01	1.18	.29	0.002	0.73	.50	0.003	
Person perception – Object-specific	1.68	.19	0.02	3.07	.09	0.01	0.57	.61	0.003	
Person perception – Language	1.68	.19	0.03	1.24	.28	0.001	1.34	.27	0.005	
Person perception – Multiple-demand	1.27	.29	0.02	7.84	.01*	0.009	1.02	.38	0.003	
Object-specific – Language	0.88	.45	0.02	0.15	.70	0.0001	0.24	.84	0.001	
Object-specific – Multiple-demand	1.61	.21	0.02	10.32	.004*	0.01	0.46	.68	0.002	
Language – Multiple-demand	1.65	.20	0.03	5.20	.03*	0.009	0.87	.45	0.004	

Table	e S9.	Comparison of w	ithin- and	l between-network	functional	connectivity	during	interactions v	with a	human ai	nd robotic agen	t across the	four runs

Pearson's correlation coefficients (Fisher z-transformation) were calculated between time courses for all possible combinations of ROIs for each within- and between-network combination per interaction and entered in a 2 (interaction: HHI, HRI) by 4 (run: 1-4) repeated-measures ANOVA for each network separately. Significant effects are in bold, \* Did not survive Bonferroni correction (p < .05 / 15);  $\eta^2_G$ : generalised eta-squared.

Network	Slope of HH	HI	Slope of HRI		Difference in slopes betw	veen HHI and HRI
	Estimates	CI (95%)	Estimates	CI (95%)	Estimates	CI (95%)
Theory-of-Mind – Theory-of-Mind	0.01	-0.01 - 0.03	0.01	-0.01 - 0.03	0.01	-0.02 - 0.03
Person perception – Person perception	0	-0.02 - 0.02	0	-0.01 - 0.02	0	-0.01 - 0.02
Object-specific – Object-specific	-0.01	-0.030.00	-0.01	-0.02 - 0.01	0.01	-0.01 - 0.02
Language – Language	0	-0.01 - 0.01	0	-0.01 - 0.01	0	-0.01 - 0.02
Multiple-demand – Multiple-demand	0	-0.01 - 0.01	0.01	-0.01 - 0.02	0	-0.01 - 0.01
Theory-of-Mind – Person perception	-0.03	-0.07 - 0.01	0.01	-0.01 - 0.03	0.01	-0.01 - 0.02
Theory-of-Mind – Object-specific	-0.02	-0.05 - 0.02	0	-0.02 - 0.01	0.01	-0.01 - 0.02
Theory-of-Mind – Language	-0.02	-0.06 - 0.01	0	-0.01 - 0.01	0.00	-0.01 - 0.02
Theory-of-Mind – Multiple-demand	-0.02	-0.05 - 0.01	0	-0.01 - 0.01	0.01	-0.01 - 0.02
Person perception – Object-specific	-0.03	-0.06 - 0.01	0	-0.02 - 0.01	0.01	-0.01 - 0.02
Person perception – Language	-0.02	-0.06 - 0.01	0	-0.01 - 0.01	0.01	-0.01 - 0.02
Person perception – Multiple-demand	-0.02	-0.05 - 0.00	0	-0.01 - 0.01	0.00	-0.01 - 0.01
Object-specific – Language	-0.01	-0.04 - 0.03	0	-0.01 - 0.01	0.00	-0.01 - 0.01
Object-specific – Multiple-demand	-0.03	-0.06 - 0.00	0	-0.01 - 0.01	0.00	-0.01 - 0.02
Language – Multiple-demand	-0.02	-0.06 - 0.01	0	-0.01 - 0.02	0.01	-0.01 - 0.02

Table S10. Estimated posterior regression coefficient for the functional connectivity Bayesian linear model

For Bayesian re-analysis, the following model was specified:  $\operatorname{corz} \sim \operatorname{run}^*$  interaction + (run | sub), with corz representing the Fisher z-transformed Pearson's correlation coefficients between the time courses for all possible combinations of ROIs for each within- and between-network combination with type of interaction (HHI, HRI) and run (1-4) as fixed effects and a random intercept and slope for participants (run | sub). The hypothesis function was used to specify the slope for the HRI: run + run:interactionhri = 0. CI (95%): 95% credibility interval



Figure S1. Comparison of the whole-brain brain analyses between the present study and the study of Rauchbauer and colleagues (2019). Reanalysis of the contrasts used in Rauchbauer and colleagues (2019) resulted in similar activation maps when contrasting the interactions with a human and robotic agent with the implicit baseline (c.f. instruction screen), human-human interactions (hhi) versus human-robot interactions (hri), and vice versa. For visual purposes contrast maps are shown. The maps from Rauchbauer and colleagues (2019) are obtained from https://neurovault.org/images/112528/.



without harmonic to noise ratio with harmonic to noise ratio

Figure S2. Comparison of the whole-brain brain analyses with and without harmonic to noise ratio as a parametric predictor. Similar effects were obtained when controlling for voice quality of the agent or general signal-to-noise using the mean harmonic to noise ratio (centred and scaled) as a parametric predictor. For visual purposes activation maps are shown with an uncorrected threshold of p < .001 (k = 10).



**Figure S3. Posterior distribution for the region-of-interest Bayesian linear model.** The following linear Bayesian regression model was specified: value ~ dynamics.d \* interaction.d + (1|sub), with value representing the beta estimates extracted for each event for each ROI averaged across network, and dynamics and type of interaction as fixed effects and a random intercept for participants (sub). Deviation coding style was used with 0.5: HHI/listening, and -0.5: HRI/speaking. Negative values for the main effects indicate a bias towards HRI or speaking, while positive values indicate a bias towards HHI or listening.



Figure S4. No effect of repeated experience on functional connectivity during interactions with a human and robotic agent. Within- and between-network z-transformed correlations are shown for the four runs and interaction type (human-human or human-robot interaction) separately. Rain cloud plots with errors bars reflecting 95% confidence intervals are used (Allen et al., 2021).

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