Social Cognitive and Affective Neuroscience, 2020, 63-73

doi: 10.1093/scan/nsaa022 Advance Access Publication Date: 12 February 2020 Original Manuscript

Theory of minds: managing mental state inferences in working memory is associated with the dorsomedial subsystem of the default network and social integration

Meghan L. Meyer, and Eleanor Collier

Department of Psychological and Brain Sciences, Dartmouth College, HB 6207 Moore Hall, Hanover, NH 03755, USA

Correspondence should be addressed to Meghan L. Meyer, Dartmouth College, HB 6207 Moore Hall, Hanover, NH 03755, USA. E-mail: Meghan.L.Meyer@Dartmouth.Edu

Abstract

OXFORD

We often interact with multiple people at a time and consider their various points of view to facilitate smooth social interaction. Yet, how our brains track multiple mental states at once, and whether skill in this domain links to social integration, remains underspecified. To fill this gap, we developed a novel social working memory paradigm in which participants manage two- or four-people's mental states in working memory, as well as control trials in which they alphabetize two- or four-people's names in working memory. In Study 1, we found that the dorsomedial subsystem of the default network shows relative increases in activity with more mental states managed in working memory. In contrast, this subsystem shows relative decreases in activity with more non-mental state information (the number of names alphabetized) managed in working memory. In Study 2, only individual differences in managing mental states in working memory, specifically on trials that posed the greatest mental state load to working memory, correlated with social integration. Collectively, these findings add further support to the hypothesis that social working memory relies on partially distinct brain systems and may be a key ingredient to success in a social world.

Key words: theory of mind; social cognition; default network; working memory

Imagine the following scenarios: a business meeting in which employees and bosses keep track of multiple colleagues' intentions; an interracial interaction in which individuals manage their stereotypes and anxieties during social inference; a gossip session in which friends tailor what is said based on one another's knowledge of the situation. These scenarios are tied by a common theme: people actively manage a great deal of social information in mind to ensure smooth social interactions. How do we pull off such complex social information processing on the fly?

The maintenance and manipulation of high-level social information (e.g. personality traits, mental states, interpersonal relationships) has been characterized as social working memory (SWM) and there is evidence that this process relies on partially distinct brain systems from those supporting non-social forms of working memory. Specifically, regions of the brain's default

Received: 9 July 2019; Revised: 8 January 2020; Accepted: 6 February 2020

© The Author(s) 2020. Published by Oxford University Press.

This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0/), which permits unrestricted reuse, distribution, and reproduction in any medium, provided the original work is properly cited.

network, which includes medial prefrontal cortex (MPFC), dorsomedial prefrontal cortex (dMPFC), precuneus/posterior cingulate (PC/PCC), tempoparietal junction (TPJ) and temporal poles (TPs), systematically increase activation when participants maintain and manipulate high-level social information in their minds (i.e. information about friends' personalities; Meyer et al., 2012; Meyer et al., 2015). In contrast, this same network is wellknown to systematically disengage during non-social forms of working memory (McKiernan et al., 2003; Anticevic et al., 2010). In fact, the default network even shows this differential relationship when participants process the same stimuli either socially or non-socially in working memory. In one study, participants completed working memory trials in which they were shown two, three or four of their own friends' names and next either ranked them along trait dimensions (e.g. who is the funniest?) or alphabetized them (e.g. whose name is first alphabetically?). Within the same participants, the default network parametrically increased activation as a function of the number of friends' traits considered in working memory even though it parametrically decreased activation as a function of the number of friends' names alphabetized in working memory (Meyer et al., 2015). Thus, the default network appears to be differentially associated with high-level social and non-social information processing in working memory.

While these findings are provocative, they generate many more questions than answers. A particularly important question to answer is whether the results generalize to multiple types of high-level social information processing managed in working memory. In fact, at first blush, it appears as though the findings may not generalize to managing mental state inferences (as opposed to traits) in working memory. Specifically, maintaining others' emotional facial expressions in working memory is associated with increased activity in the lateral frontoparietal network (Smith et al., 2017), but relatively suppressed activity in the default network (Xin & Lei, 2015). However, in these paradigms, participants are asked to judge whether a facial expression of a 'basic emotion' (e.g. fear, happiness; Eckman, 1999) matches the basic emotion expressed in a previous trial. Thus, task performance does not necessitate managing mental state inferences in working memory. Mental state inference refers to the process of determining what other people think and feel (Ames, 2004; Mitchell, 2009); however, in these past studies, only the physical facial expressions need to be maintained to match expressions. As a result, it remains unclear whether the default network may increase activity in response to managing mental state inferences in working memory.

Another important question to answer is whether SWM skills relate to social integration. It has been suggested that just as non-SWM skills relate to academic success (Conway et al., 2002; St. Clair-Thompson and Gathercole, 2006; Rohde and Thompson, 2007), SWM skills may relate to interpersonal success (Meyer et al., 2012; Meyer et al., 2015; Krol et al., 2018; Smith et al., 2018a; Mikels and Reuter-Lorenz, 2019). Indeed, individuals who can accurately manage more high-level social information in mind should also make the most accurate predictions about others, and critically, feeling accurately understood by peers promotes friendship with them (Reis and Shaver, 1988; Cross et al., 2000). Yet, pinpointing whether SWM accuracy relates to social integration has been challenging, in part, because responses to the SWM paradigm previously used to assess the relationship with social integration are subjective. That is, past work has linked social integration to SWM performance on the trait ranking paradigm, in which participants rank their friends from 'most-to-least' along trait dimensions over a delay period (Krol et al., 2018).

However, when a participant indicates that one of their friends is funnier than the other on a given SWM trial, it is unclear whether or not they answered accurately. What is needed is a SWM paradigm in which task accuracy can be computed more objectively, so that individual differences in SWM capacity can be linked to differences in social integration.

The goal of the present research was to create a new SWM paradigm to fill these gaps. In this paradigm, participants first watch a video montage, which introduces a social network of characters with different interpersonal relationships (i.e. friends, lovers, competitors and enemies). Participants next complete a SWM task in which they are asked to determine how a given character would feel, based on other characters' feelings. Critically, the number of characters shown varies from trial to trial, with half of the trials showing two characters and the other half showing four characters. This allows for the isolation of neural activity and task performance as a function of SWM load, above and beyond mental state inference per se. Because interpersonal relationships tend to create known emotional contingencies between people (e.g. people tend to be sad to learn their friends are unhappy but often delight to hear their enemies are unhappy (Cikara et al., 2011)), we were able to measure SWM task accuracy more objectively. The paradigm also included non-SWM trials that were matched on a number of features (e.g. task difficulty), but that did not induce mental state inference. With this paradigm, we were able to test whether considering multiple mental state inferences in working memory differentially engages the default network (Study 1) and whether SWM task accuracy uniquely predicts social integration (Study 2).

Study 1

Participants

Thirty-seven participants (13 male, 23 female, 1 unspecified; mean age = 29.32 years, SD = 11.32; 65% Caucasian, 19% Asian, 5% Hispanic or Latino, 11% other) completed Study 1 for course credit or monetary payment (\$20/h). Participants were eligible to participate if they did not have any metal in their body, were not claustrophobic and were right-handed. Sample size was determined by available funding, and a priori power analysis in G*Power 3.1 (Faul *et al.*, 2007) showed that this sample size is estimated to provide 99% power to detect a medium withinsubject working memory type × load level interaction effect (i.e. $np^2 = .09$). Participants provided informed consent in accordance with the Dartmouth College Institutional Review Board (IRB).

Procedure

Social working memory paradigm. Directly before fMRI scanning, participants first watched a montage introducing them to characters from the television show *Orange* is the New Black. A professional television editor developed the montage to help ensure that it effectively communicated the interpersonal relationships between the characters in an engaging format. Five characters were introduced and relationships between the characters included friends, enemies, lovers and competitors. Before running the study, a sample of pilot subjects watched the montage and next described the relationships between the characters (N = 19). All participants accurately described the relationships, suggesting that the relationship dynamics are well conveyed by the montage.

In Study 1, after watching the montage, participants indicated whether or not they had previously seen the television



Fig. 1. Social working memory (SWM) paradigm. For SWM trials (Panels A–B), participants determine how the target character (pink box) would feel, based on the other characters' feelings. The red thumbs-down sign indicates that a character is feeling negatively, whereas the green thumbs-up sign indicates that a character is feeling positively. For non-SWM trials (Panels C–D), participants alphabetize the characters' names, based on whose name anchors the alphabet line (which is indicated by the anchor sign). Panels A and C show two-load working memory trials and Panels B and D show four-load working memory trials. Participants encode the initial stimuli for 4 s, which is followed by a 4-s delay period. Participants next have up to 3 s to make their response.

show Orange is the New Black. This allowed us to subsequently examine whether task performance varies as a function of familiarity with the characters in the montage. Participants next received instructions regarding how to complete the SWM and non-SWM trials and completed six, unique practice trials (two per condition) on a laptop so that they were familiar with the task prior to scanning.

While undergoing fMRI scanning, participants completed the primary task (Figure 1), which employs a 2 (working memory type: SWM vs. non-SWM) \times 2 (working memory load: two vs. four) experimental design. Participants completed 36 randomly presented trials (9 per condition). For two-load SWM trials, participants see two characters on the screen (4 s). The mental state of one character is provided, indicated by a thumbs-up (positive mental state) or thumbs-down (negative mental state). The mental state of the other 'target' character is not provided. Next, participants are instructed to consider how the target character would feel, based on the other character's given mental state and relationship with the target character during a delay period (4 s). For example, if the two characters are friends and the character with the shown mental state is feeling positive, then the participant would reason that the other character, who likes their friend, would feel positive. After making the rating on a button box, the trial advanced to jittered fixation for a randomly generated duration between the range of 1.55-4.47 s (M = 2.94 s).

Four-load SWM trials have a similar format. The key difference is that for these trials participants are shown three characters with mental states and one target character with no mental state. During the subsequent delay period, participants reason how the target character would feel, based on the other characters' feelings. Thus, two-load and four-load trials are identical, except that the four-load trials afford greater SWM load. To further help determine task accuracy objectively, participants were instructed to reason about the mental states serially and 'from left to right.' That is, in the example shown in Figure 1B, participants would reason about Piper's mental state in the following order: Larry's influence on her mental state (she would be unhappy if Larry, her fiancé, was unhappy; thus a rating of 1), her happiness would be taken up a notch if she find out that her enemy Tiffany was unhappy (now a rating of 2) and up another notch if she found out Claire, her girlfriend, is happy (now a rating of 3, the final, correct answer for this trial). In this way, mental state inferences proceed one at a time, and each additional character's mental state is considered independently of one another. For all SWM trials, the target character always had a relationship with the other shown characters, ensuring that the target's mental state would be influenced by each of the other characters' mental states.

Non-SWM trials have the same format, except here participants alphabetize the characters' names over a delay period, also based on a (non-social) relationship between stimuli (i.e. the relationship between the position of letters on the alphabet line). Participants see a set of characters and an anchor sign is shown under one character (4 s). Participants are told that the anchor indicates that for a given trial, the alphabet line should start with the first letter of the character's name with the anchor. For example, if the anchor appeared under the character named Piper, 'P' should be treated as the first letter of the alphabet, and once reaching 'Z' the alphabet line would wrap around so that 'O' would become the last letter of the alphabet. During the delay period, participants alphabetize the characters' names based on this relational rule between stimuli. To ensure that the answer could vary for two-load trials, participants were shown the character with the anchor for half of the probe-responses and the target character for the other half of the probe-responses. As is the case in the SWM trials, the four-character trials were identical to the two-character trials, but simply afforded a greater working memory load. In their probe response, participants indicated the alphabetical position of the shown character. This control condition was chosen to be consistent with past, non-SWM research. Many working memory studies have parameterized the amount of alphabetizing over a delay period to assess the brain systems associated with working memory (e.g. D'Esposito et al., 1999; Postle et al., 1999; Postle et al., 2006; Fougnie and Marois, 2007; Maniscalco and Lau, 2015). Creating this alphabetizing condition allowed us to be consistent with prior research, while also having participants reason based on (alphabetical) relationships between encoded stimuli.

fMRI scanning. Scanning was conducted with a Siemens Trio 3 T Prisma. Participants first completed a T2-weighted structural scan acquired coplanar with the functional images and with the following parameters: TR = 2300 ms, TE = 2.32 ms, 0.9 mm slice thickness, FOV = 24 cm, matrix = 256×256 , flip angle = 8°. During this time, participants watched the *Orange* is the New Black montage for a second time to refresh their memory of the relationships between characters. The task was completed during two functional runs using an EPI gradient-echo sequence with the following parameters: TR = 1000 ms, TE = 30 ms, 2.5 mm slice thickness, FOV = 24 cm, matrix = 96×96 and flip angle = 59° .

fMRI data analysis. Functional brain imaging data was first reoriented in SPM8 (Wellcome Department of Cognitive Neurology, Institute for Neurology, London, UK) and skull-stripped with Brain Extraction Tool (BET; Smith, 2002) in FSL. All subsequent fMRI analyses were performed with SPM8. Preprocessing steps included spatial realignment to correct for head motion, normalization into a standard stereotactic space as defined by the Montreal Neurological Institute, and spatial smoothing using an 8-mm Gaussian kernel, full width at half-maximum.

For each subject, neural activity was modeled for each condition using a general linear model. Because we observed differences in reaction time (RT) across our SWM and non-SWM conditions (see Results section below), we also ran models that included, for each condition, RT as a parametric modulator. These models were run to ensure that clusters that varied between content (SWM vs. non-SWM) and load level (two-load vs. four-load) were not conflated with differences in RT across these conditions. Participants' first-level models included a regressor for each condition of interest (convolved with the hemodynamic response function (HRF) vs. implicit baseline (i.e. SWM two-load vs. implicit baseline; SWM four-load vs. implicit baseline; non-SWM two-load vs. implicit baseline; non-SWM four-load vs. implicit baseline), as well as six motion regressors for each of the motion parameters from image realignment.

Whole-brain analyses. Next, we used a flexible factorial design to identify clusters of neural activity across subjects that are differentially associated with four-load (vs. two-load) SWM trials relative to four-load (vs. two-load) non-SWM trials. For completeness, we also tested the effect of working memory content (SWM vs. non-SWM) and load level (two characters vs. four characters). Whole-brain results were voxelwise thresholded at P < .001, with a cluster-extent threshold corrected for familywise error (FWE) rate, using a cluster-defining FWE threshold of P < .001, as determined by SPM. This approach corresponded with a cluster-extent threshold of 352 voxels. Predefined default network analyses. We followed up our wholebrain interaction results with analyses performed with predefined default network subsystems. We used the default network subsystems defined by Yeo *et al.* (2011), which were generated from 1000 participants' resting state scans, and are thus highly reliable. This approach allowed us to assess whether (i) our results are replicated in a formally defined default network and (ii) the interaction we observed in the dorsomedial subsystem in the whole-brain analysis was indeed reflected by relatively greater activity to SWM four-load (vs. two-load) trials and relatively less activity to non-SWM four-load (vs. two-load) trials, using clusters defined independently of the whole-brain results.

We extracted parameter estimates from each condition (vs. implicit baseline) for each of the three default network subsystems identified by Yeo et al. (2011). The subsystems include the dorsomedial subsystem (dorsomedial prefrontal cortex (DMPFC), temporoparietal junction (TPJ), middle temporal gyrus (MTG) extending into TP and inferior frontal gyrus (IFG)), core subsystem (medial prefrontal cortex (MPFC), posterior cingulate/precuneus (PC/PCC), posterior inferior parietal lobule (PIPL)) and medial temporal lobe subsystem (hippocampal formation, retrosplenial cortex and dorsal posterior inferior parietal lobule (dPIPL); Figure 4A). We then tested the interaction and appropriate follow-up t-tests for each network subsystem, separately. Supplementary Figure 1 shows the overlap between the clusters observed in the whole-brain interaction and those identified by Yeo et al. (2011) as part of the dorsomedial subsystem, which includes substantial overlap in DMPFC, TPJ and TPs.

Results

Task performance. Task accuracy was defined as the percentage of correctly answered trials for each condition. There was no significant interaction between working memory content (SWM vs. non-SWM) and load level (two-load vs. four-load; F(1,36) = .05, P = .820, $np^2 = .001$; Figure 2A), suggesting that task performance was equivalent between the SWM and non-SWM trials. Consistent with this, there was no main effect of working memory content (SWM vs. non-SWM; F(1,36) = 2.97, P = .094, $np^2 = .08$). In contrast, there was a significant main effect of load level $(F(1,36) = 81.90, P < .0001, np^2 = .70)$ such that four-load trials were more challenging (M = 55%, SD = 24%) than two-load trials (M = 82%, SD = 17%). Task accuracy did not statistically differ between participants who had previously seen Orange is the New Black (N = 15) relative to participants who had not seen the show (N = 22; P's across load level comparisons>.177, Cohen's d's < .42; see Supplementary Table 1). Collectively, these results suggest that task accuracy was equivalent for the SWM and non-SWM trials, that the four-load manipulation successfully taxes SWM and non-SWM and that prior exposure to the television show Orange is the New Black does not moderate task performance.

RT was defined as the speed (in seconds) with which participants made correct responses. Although participants' accuracy was equivalent on the SWM and non-SWM trials, RT varied across conditions. There was a significant content (SWM vs. non-SWM) × load level (two-load vs. four-load) interaction (F(1,36) = 31.21, P < .0001, $np^2 = .48$), such that the decrement in speed for the four-load vs. two-load trials was larger for non-SWM trials (two-load M = .98, SD = .26; four-load M = 1.74, SD = .36) than SWM trials (two-load M = .88, SD = .32; four-load M = 1.11, SD = .65). There was also a main effect of content (SWM vs. non-SWM, F(1,36) = 28.56, P < .0001, $np^2 = .46$), such that participants responded more quickly to SWM trials (M = 1.00, SD = .47) than non-SWM trials (M = 1.34, SD = .24). As is the case for task

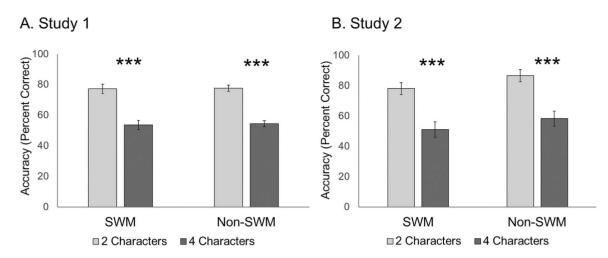


Fig. 2. Task accuracy in Study 1 (Panel A) and Study 2 (Panel B). In both studies, accuracy was matched for the SWM and non-SWM trials as a function of load level. Additionally, four-load trials were significantly more challenging than two-load trials. Asterisks (***) indicate P < .001.

accuracy, there was also a main effect of load level (two-load vs. four-load, F(1,36) = 131.32, P < .0001, $np^2 = .79$) such that participants responded more quickly to two-load (M = .93, SD = .25) than four-load trials (M = 1.40, SD = .41). RT did not statistically differ between subjects who had previously seen the television show *Orange* is the New Black compared to those who had not (P's across load level comparisons>.481; Cohen's d's < .26).

Whole-brain results. The primary goal of the neuroimaging analyses was to test whether default network regions are differentially associated with SWM and non-SWM load. To test this, we searched for clusters of activity more strongly associated with four-load (vs. two-load) SWM trials relative to four-load (vs. two-load) non-SWM trials. This analysis revealed clusters in the dorsomedial subsystem of the default network, including DMPFC, TPJ, MTG extending into TP and IFG; Figure 3A, Table 1). A cluster outside of the default network in the putamen also emerged in this analysis. The reverse contrast (greater activity for four-load (vs. two-load) non-SWM trials relative to fourload (vs. two-load) SWM trials) showed clusters of activation in supplementary motor area and precentral gyrus (Table 1).

Collapsing across load level, we found that multiple regions in the default network (DMPFC, VMPFC, left PC/PCC, ITPJ and bilateral TP) showed greater activity in response to the SWM vs. non-SWM trials (Supplementary Figure 2, Supplementary Table 2). These results further support the idea that the default network is associated with SWM, and mental state inference more generally. Collapsing across working memory content, and replicating past working memory research, clusters in DLPFC and the precuneus extending into bilateral inferior parietal lobule, showed greater activity in response to four-load vs. two-load trials (Supplementary Figure 2, Supplementary Table 2). For the interested reader, clusters that emerged in the comparison of non-SWM vs. SWM included supplementary motor area, anterior insula, palladium, precuneus and the cerebellum (Supplementary Figure 2, Supplementary Table 2). Only MPFC and superior temporal lobe showed relatively greater activity to two-load vs. fourload trials (Supplementary Figure 2, Supplementary Table 2). Follow-up analyses showed that all clusters that differentiated SWM and non-SWM trials remained significant when analyses included, for each condition, RT as a parametric modulator (Supplementary Figure 3), suggesting observed differences in neural activity cannot be explained by differences in RT between SWM and non-SWM trials.

Predefined default network subsystem results. Next, we assessed whether our whole-brain results replicate in a predefined set of default network subsystems. Specifically, we used the default network subsystems identified by Yeo et al. (2011), which includes dissociable regions for the dorsomedial subsystem, core subsystem and MTL subsystem (Figure 4A). We again observed a significant content (SWM vs. non-SWM) \times load level (two-load vs. four-load) interaction in the default network's dorsomedial subsystem (F(1, 36) = 11.356, P = .002, np^2 = .240). Follow-up paired samples t-tests confirmed that the dorsomedial network was significantly more active during SWM four-load vs. SWM twoload trials (t(36) = 2.273, P = .029). The dorsomedial subsystem was also significantly less active during non-SWM four-load vs. non-SWM two-load trials (t(36) = 2.68, P = .011). There was also a main effect of content (F(1,36) = 4.907, P = .033, $np^2 = .120$) such that the dorsomedial subsystem was relatively more active during SWM (M = .29, SD = .24) vs. non-SWM trials (M = .11, SD = .33). There was no main effect of load level (F(1, 36) = .281, $P = .599, np^2 = .008$).

The core subsystem did not show a significant interaction F(1, 36) = 1.997, P = .166, $np^2 = .053$), nor a main effect of content (F(1, 36) = 1.682, P = .203, $np^2 = .045$) or load level (F(1, 36) = .202, P = .656, $np^2 = .006$). Although the MTL subsystem showed a significant content \times load level interaction (F(1, 36) = 5.578, P = .024, $np^2 = .134$), this effect was driven by differences in activity in the non-SWM four-load vs. two-load trials (t(36) = 2.110, P = .042), but not SWM four-load vs. two-load trials (t(36) = 0.802, P = .427). The MTL subsystem also did not show a main effect of content (F(1, 36) = 1.070, P = .308, $np^2 = .029$) or load level (F(1,36) = .601, P = .443, $np^2 = .016$). Subsystem results are displayed in Figure 4B.

The goal of Study 1 was to test whether default network regions show relative increases in activity as a function of the number of mental states managed in working memory. The dorsomedial subsystem of the default network showed relative gains in activation as a function of the number of mental states managed in working memory, despite showing relative decreases in activity as a function of the number of names alphabetized in working memory. Critically, SWM and non-SWM trials were matched on task accuracy and these neural patterns persisted when controlling for differences in RT across conditions. Collectively, these results suggest that engaging default network regions during SWM generalizes to managing mental state inferences in working memory. Because our task measures Table 1. Clusters of neural activity more strongly associated with four-load (vs. two-load) SWM trials relative to four-load (vs. two-load) non-SWM trials (top) and four-load (vs. two-load) non-SWM trials relative to four-load (vs. two-load) SWM trials (bottom). Columns x, y, and z refer to cluster coordinates, t refers to the t-statistic associated with the cluster, and k refers to the cluster extent (number of voxels)

	Region	x	У	Z	t	k
Interaction: Social (Four-Load 4 vs. Two-Load) vs.						
Non-Social (Four-Load 4 vs. Two-Load)						
······	Dorsomedial Prefrontal Cortex (DMPFC)	12	36	54	6.04	1598
	DMPFC	-6	34	56	5.6	
	DMPFC	-4	44	48	5.68	
	Tempoparietal Junction (TPJ)	-52	-64	40	5.67	576
	ITPJ	-48	-58	30	4.49	
	ITPJ	-56	-64	26	4.30	
	Inferior Frontal Gyrus (IFG)	48	42	-14	5.94	523
	IFG	46	32	-12	5.23	
	IFG	46	44	-4	4.8	
	Putamen	32	-12	-2	5.87	921
	Posterior Insula	40	-4	-2	4.52	
	Posterior Insula	34	-24	-2	4.55	
	Precentral Gyrus	38	-20	52	6.32	3075
	Middle Temporal Gyrus	62	-42	2	6.26	
	rTPJ	52	-60	44	5.71	
Interaction: Social (Four-Load 4 vs. Two-Load) vs.						
Non-Social (Four-Load 4 vs. Two-Load)	Supplementary Motor Area	-4	0	64	5.83	1698
	Precentral Gyrus	-36	-24	52	4.69	

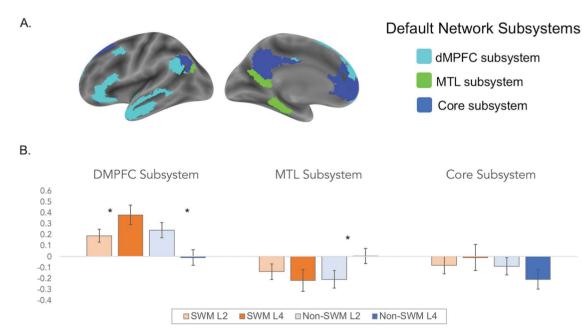


Fig. 3. Neuroimaging results for the interaction contrast testing for activity associated with four-load (vs. two-load) SWM trials relative to four-load (vs. two-load) non-SWM trials. Panel A shows clusters in the dorsomedial subsystem of the default network (DMPFC, TPJ, MTG, TP, IFG). Panel B shows parameter estimates from each of these clusters for each condition.

SWM performance objectively and given past suggestions that SWM skill may relate to social success (Meyer et al., 2012; Meyer et al., 2015; Krol et al., 2018; Smith et al., 2018; Mikels et al., 2019), our next goal addressed in Study 2 was to assess whether individual differences in SWM preferentially relate to individual differences in social integration.

Study 2

Participants

Eighty-eight participants (males=44, females=44; average age=23.53 years, SD=4.49.66% Caucasian, 16% Asian, 9% Black

or African American, 7% Hispanic or Latino, 2% Other) completed Study 2. Given research showing that social network size decreases during the years following young adulthood (English and Carstensen, 2014), to avoid age confounds in our results we constrained recruitment to participants between the ages of 18 and 29. Sample size was determined by a power analysis in G*Power 3.1 (Faul *et al.*, 2007), which showed that at least 85 participants were needed to allow for 80% power to detect a medium effect size (i.e. Pearson r = .30) when examining the correlation between task performance and social network size. Participants completed the study for course credit or monetary payment (\$6).

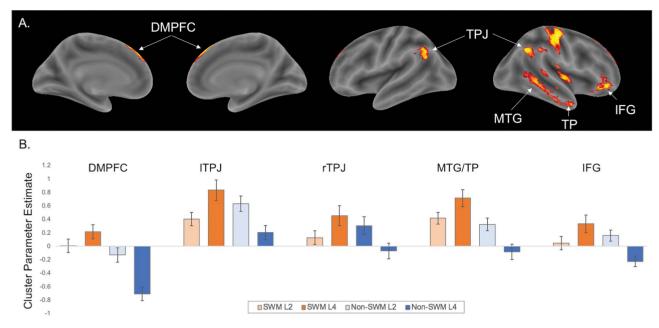


Fig. 4. Results from the default network subsystem analyses. Panel A shows the default subsystems, which were generated from the resting state data of 1000 participants (Yeo *et al.*, 2011). Panel B shows activation in the default network subsystems across the social and non-social working memory trial types.

Procedure

Social working memory paradigm. Participants completed the SWM paradigm developed for Study 1. Participants watched the Orange is the New Black montage and indicated whether or not they had seen the television show in the past. Next, they received task instructions, completed practice trials and finally performed the experimental task consisting of 36 randomly presented trials (9 per condition).

Social integration. To assess social integration, we used a questionnaire developed by Stiller and Dunbar (2007) which measures the number of social network members interacted with over the past 7 days. In this measure, participants are asked to list all of the people with whom they had personal contact or communication during the past 7 days, excluding (i) work colleagues seen only in a work environment (unless they considered them genuine friends), (ii) contacts with professionals for appointments (such as doctors) and (iii) other casual acquaintances (e.g. brief encounters in a shop).

Results

Task performance. We replicated the task accuracy findings observed in Study 1. There was neither a significant content (SWM vs. non-SWM) × load level (two-load vs. four-load) interaction (F(1, 87) = 1.67, P = .200, np^2 = .02), nor a significant main effect of content (SWM vs. non-SWM; F(1, 87) = 1.36, P = .248, $np^2 = .02$; Figure 2B). These findings again suggest that the SWM and non-SWM trials are matched on accuracy. There was a significant main effect of load level (two-load vs. fourload; F(1, 87) = 139.32, P < .0001, np² = .61) such that four-load trials (M = 75%, SD = 19%) were more challenging than two-load trials (M=54%, SD=20%). As in Study 1, task accuracy did not statistically differ between participants who had previously seen Orange is the New Black (N = 32) relative to participants who had not (N = 39; P's > .407, Cohen's d's < .19; Supplementary Table 1). Collectively, these results again suggest that difficulty level (in terms of accuracy) was equivalent for the SWM and

non-SWM trials, that the four-load manipulation successfully taxes working memory resources and that previous exposure to the television show *Orange* is the *New Black* does not moderate task performance.

We also replicated the RT results observed in Study 1. There was a significant content (SWM vs. non-SWM) × load level (twoload vs. four-load) interaction (F(1, 84) = 48.28, P < .0001, np² = .37), such that the decrement in speed for the four-load vs. twoload trials was larger for non-SWM trials (two-load M = 1.04, SD = .41; four load M = 1.67, SD = .37) than SWM trials (two-load M = .73, SD = .29; four-load M = .92, SD = .39). There was a main effect of content (SWM vs. non-SWM, F(1, 84) = 255.47, P < .0001, np^2 = .75), such that participants responded more quickly to SWM (M = .82, SD = .29) than non-SWM trials (M = 1.35, SD = .26). As is the case for task accuracy, there was also a main effect of load level (two-load vs. four-load, F(1, 84) = 185.48, P < .0001, $np^2 = .69$) such that participants responded more quickly to twoload (M = .89, SD = .24) vs. four-load trials (M = 1.30, SD = .29). RT did not statistically differ between subjects who had previously seen Orange is the New Black compared to those who had not (P's across load level comparisons>.105; Cohen's d's < .40).

Relationship between SWM and social integration. Interaction with social network members significantly correlated with SWM accuracy on the two-load trials (r = .21, P = .045) and four-load trials (r = .31, P = .003, Figure 5A). A regression analysis demonstrated that the relationship between social network member interaction and four-load SWM accuracy remained significant when controlling (i.e. adding as a covariate) two-load SWM accuracy (SWM L4 β = .30, t = 2.23, P = .028; SWM L2 β = .03, t = .21, P = .834). Thus, SWM capacity, above and beyond mental state inference in general, is associated with social network integration. In contrast, interaction with social network members did not significantly correlate with non-SWM accuracy on the twoload (r = .05, P = .643, Supplementary Figure 4) or four-load trials (r = .17, P = .108, Supplementary Figure 4). That said, it is noteworthy that the relationship between social integration and non-SWM four-load trials, which can be considered the non-SWM 'yoked control condition' for the SWM four-load trials, showed

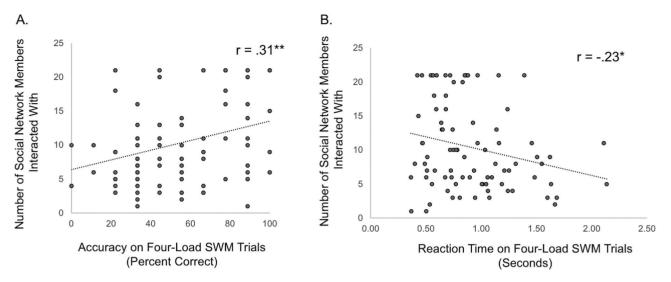


Fig. 5. Relationship between social working memory (SWM) task performance on four-load trials and the number of social network members interacted with over a seven-day period. Panel A shows results for task accuracy and Panel B shows results for reaction time (RT). ** indicates P < .005 and * indicates P < .05.

a trend in the positive direction. We therefore ran a followup regression analysis to investigate whether the relationship between social network integration and SWM four-load accuracy persists when controlling for non-SWM four-load accuracy. This was indeed the case (four-load SWM accuracy β = .28, t = 2.62, *P* = .010) and is consistent with the primary results showing that SWM is associated with social integration.

RT analyses paralleled those observed for task accuracy. Individuals who interacted with the most social network members were also faster to make a correct response to SWM four-load trials (r = -.23, P = .032, Figure 5B) and marginally SWM two-load trials (r = -.19, P = .073). In contrast, interaction with social network members was not significantly related to RT on non-SWM four-load trials (r = -.09, P = .393, Supplementary Figure 4) nor two-load trials (r = -.17, P = .107, Supplementary Figure 4).

Discussion

How do we manage mental state inferences in working memory? Here we found that the dorsomedial subsystem of the default network shows relative increases in activity as a function of the number of mental states considered in working memory. In contrast, these regions showed relative decreases in activity as a function of non-mental state information (the number of names alphabetized) considered in working memory. Such findings are consistent with the suggestion that the default network is differentially associated with SWM and non-SWM (Meyer *et al.*, 2012, 2015). Moreover, only individual differences in SWM task performance, specifically on trials that posed the greatest SWM load, correlated with social integration. Thus, SWM capacity, above and beyond mental state inference and non-SWM capacity more generally, may help us navigate everyday social life.

Although the default network comprises multiple subsystems, here we found that only the dorsomedial subsystem showed relative increases in activity with the number of mental states managed in working memory. These findings are consistent with past research on mental state inference, which consistently implicate TPJ and DMPFC, key nodes in the dorsomedial subsystem, in assessing others' minds (Gallagher et al., 2000; Dodell-Feder et al., 2011; Schurz et al., 2014). Relevant to understanding the role of different default network regions in SWM, graph analytic methods show that the default network is comprised of three subsystems: the core subsystem comprising MPFC, PC/PCC and PIPL; the medial temporal lobe subsystem comprising hippocampal formation, retrosplenial cortex and dPIPL; and the dorsomedial subsystem comprising DMPFC, TPJ and MTG extending into TPs (Yeo *et al.*, 2011). Thus, it is possible that different default subsystems are preferentially associated with maintaining and manipulating different types of social information in working memory.

For example, MPFC and PC/PCC are consistently implicated in self-referential processing, as well as thinking about close others, such as friends and family (Krienen et al., 2010; Denny et al., 2012). Previous SWM research found that MPFC increased activity as a function of the number of friends' traits considered in working memory (Meyer et al., 2012, 2015). In contrast, here we observed that MPFC was less active in response to the higher (vs. lower) social and non-social working memory load trials (i.e. the comparison of four-load vs. two-load trials, collapsed across SWM and non-SWM trials). Our findings suggest that MPFC disengages in response to working memory load when managing social information about strangers, unlike when managing social information about close others. To further clarify the role of the default network subsystems in working memory, it may be helpful for future research to parameterize working memory load across self, close other and stranger targets. Interestingly, although some research has implicated MPFC in managing one's own emotional responses to affective stimuli in working memory (Waugh et al., 2014; Smith et al., 2018b; Smith et al., 2018c), to our knowledge no research has investigated the brain basis of managing self-concept information (e.g. beliefs about one's personality) in working memory. This gap is surprising given social psychological theory suggesting that managing selfconcept information in working memory is important for selfregulation and fulfilling interpersonal relationships (Markus and Wurf, 1986).

The observation that the dorsomedial subsystem preferentially supports mental states in working memory provides new insight into the functional properties of this system and its role in theory of mind (TOM), or the capacity to represent others' mental states. The classic paradigm used to assess TOM is the false-belief test (Wimmer and Perner, 1983; Dufour *et al.*, 2013). In this paradigm, participants first observe Person A witness Person B place an item in one of two boxes. Person A leaves the room and Person B moves the item to the other box. Person A returns to the room, and the participant is asked which box Person A believes the item is in. Although not typically framed as a working memory task, determining a correct answer to false belief tasks requires working memory; participants must maintain Person A's belief over time to answer correctly. Thus, an alternative interpretation of TPJ and DMPFC activity in response to false-belief tasks is that these regions support the momentary maintenance of others' mental states in working memory. One impediment to this interpretation from past work is that traditional false-belief tasks do not parameterize working memory load, making it difficult to disentangle whether these regions are important for mental state inference generally versus maintaining mental state inferences in working memory specifically. Because our two-load and four-load SWM trials required mental state inference, but only the four-load trials increased mental state load in working memory, our results provide helpful insight: TPJ and DMPFC show relative increases in activation as a function of the number of mental state demands to working memory, above and beyond mental state inference per se. These findings are consistent with another recent study finding that the TPJ shows greater activity when making two false belief inferences versus one false belief inference (Özdem et al., 2019) and suggest that activity in these regions during false belief tasks may reflect a (social) working memory process.

Our paradigm is also useful for assessing TOM capacity in adults, who traditionally perform at ceiling on most false-belief tasks (Fletcher et al., 1995; Brunet et al., 2000; Walter et al., 2004). TOM research focuses heavily on when children acquire the capacity to infer mental states (e.g. Wellman, 1992) and how these mechanisms are altered in populations who struggle to understand others, such as individuals with an autism spectrum disorder (ASD; e.g. Baron-Cohen, 2000). However, far less research examines the upper bounds of TOM expertise, perhaps in part because there are little-to-no paradigms that effectively create variability in healthy adult performance. In contrast, participants in Studies 1 and 2 did not perform at ceiling on our tasks, and individual differences in task performance related to realworld social integration. Thus, our paradigm may be an effective way to begin to unpack individual differences in adult TOM and its link to social prowess.

In contrast to our brain imaging findings, other research on managing other people's mental states in working memory has found increased activity in the lateral frontoparietal network typically associated with working memory (Smith et al., 2017), but relatively suppressed activity in the default network when participants track others' emotions (Xin & Lei, 2015). Critically, however, this past work uses paradigms in which participants maintain the external facial features of (basic) emotional expressions in working memory. For example, participants are asked to indicate whether a given person's emotional expression matches the emotional expression of the person shown two trials prior (Xin and Lei, 2015). In these paradigms, participants do not need to engage in high-level mental state inference (i.e. 'what might he/she be feeling?'); maintaining visual features of facial expression is sufficient. In our paradigm, the facial expressions of the individuals in the photographs are held constant from trial to trial, but participants must make and manage mental state inferences in working memory. In everyday life, these two processes often unfold in tandem (e.g. observe a facial expression and make a high-level social inference). To gain a more holistic understanding of social working memory

processes in everyday life, an interesting direction for future work may be to assess how external (e.g. facial expressions) and internal (e.g. social inferences) sources of information are integrated in working memory.

The behavioral results from Study 2 complement and extend past paradigms used to assess whether SWM skills relate to social network integration. Past work has shown that the ability to manage people's personality traits in working memory (Krol et al., 2018) and maintain information about characters' mental states from a story (Stiller and Dunbar, 2007) correlate with the number of social network members interacted with over the course of a week. The proposed interpretation of these results is that skill in maintaining and manipulating people's internal characteristics in working memory may help us cultivate social networks. However, two limitations impede this interpretation. The first study used the SWM task in which participants rank a set of friends along trait dimensions 'from most-to-least' in working memory and determine accuracy by comparing participants' answers on a given SWM trial to their ratings of their friends' traits provided 2 weeks earlier. Thus, task accuracy is computed with subjective assessments of friends' traits, and variables such as whether a new event changed a participant's perception of their friend(s) confound the results. The finding that the ability to remember characters' mental states from a story correlates with social network integration confounds other variables with task performance, most notably reading comprehension. The paradigm developed here overcomes these limitations, as task accuracy is computed objectively and does not rely on reading comprehension (all stimuli are presented visually). Moreover, we found that task accuracy specifically on the four-load SWM trials, above and beyond task accuracy on the two-load SWM trials, uniquely predicts the number of social network members interacted with over a week. This finding further suggests that the ability to maintain and manipulate mental states in working memory, above and beyond mental state inference per se, is associated with greater social network integration. Future work may reveal the causal direction of this relationship, assessing whether SWM skills prospectively predict social network size, whether cultivating larger social networks builds SWM capacity, or both.

Limitations. While our results provide helpful insight into the brain basis of SWM and its connection to social integration, they are not without limitations. First, although we were successful in developing a paradigm that computes SWM task accuracy more objectively than past tasks, it is noteworthy that our computation of accuracy assumes that all relationship partners' mental states equally impact one another. However, in real life, different relationship partners may impact another person's mental state to different degrees. For example, perhaps your romantic partner's mental states have a larger impact than your friend's mental states in determining your own. An interesting direction for future research will be to model the varying weights of emotional contingencies between people to more naturalistically assess how the social brain makes such inferences.

Second, although the SWM and non-SWM trials differ in mental state inference, they also differ in the extent to which they rely on new, episodic information acquired from watching the Orange is the New Black video montage. Given that the MTL default network subsystem associated with episodic memory (Squire and Zola-Morgan, 1991) did not distinguish SWM from non-SWM trials, we suspect this difference is not driving our results. Nonetheless, future work may test this possibility more directly, for example by comparing SWM responses to managing new vs. old social information. Related to SWM and non-SWM task differences, these trials also differed in RT, suggesting the conditions—despite being well-matched on accuracy—may still vary in difficulty. On the one hand, this difference in RT could suggest that the greater default network activity associated with SWM reflects this network's involvement in less vs. more effortful cognitive processing. On the other hand, this alternative explanation is unlikely, given that our findings persisted even when we controlled for RT on a trial-by-trial basis.

Third, our measure of social network integration relies on self-reported social interaction. Although this measure is commonly used in social cognition research (Roberts *et al.*, 2008; Lewis *et al.*, 2011; Heleven and Van Overwalle, 2016; Heleven and Overwalle, 2019a; Heleven and Overwalle, 2019b), it is imperfect. For example, a person might not remember all of their social interaction partners when probed 1 week later. Moving forward, future research may replicate and extend these results with other assessments of social network integration, such as those used in formal social network analyses (Parkinson *et al.*, 2017).

Conclusion

In summary, we examined how people manage mental state inferences in working memory and whether skill in this domain relates to real-world social integration. Study 1 demonstrated that the dorsomedial subsystem of the default network shows relative increases in activity as a function of the number of mental states considered in working memory, despite relative decreases in activity as a function of the number of nonmental state information considered in working memory. Study 2 showed that skill in managing more mental states in working memory preferentially correlates with social integration. These findings add further support to the idea that SWM relies on partially distinct brain systems and may be critical to interpersonal success.

Supplementary data

Supplementary data are available at SCAN online.

Acknowledgements

We would like to thank Joseph Mastromonaco for creating the *Orange* is the New Black montage and Matthew Lieberman for his suggestions on the experimental design. We would also like to thank Thuy-Vy Nguyen and Remy Borinsky for their assistance with data collection and analysis for Study 2.

References

- Ames, D.R. (2004). Inside the mind reader's tool kit: projection and stereotyping in mental state inference. *Journal of Personality and Social Psychology*, **87**(3), 340–52.
- Anticevic, A., Repovs, G., Shulman, G.L., Barch, D.M. (2010). When less is more: TPJ and default network deactivation during encoding predicts working memory performance. *NeuroImage*, 49(3), 2638–48.
- Baron-Cohen, S. (2000). Theory of mind and autism: a fifteen year review. In: Cohen, D.J., editor. Understanding Other Minds: Perspectives from Developmental Cognitive Neuroscience, Oxford: Oxford University Press, pp. 1–18.
- Brunet, E., Sarfati, Y., Hardy-Bayle, M.C., Decety, J. (2000). A PET investigation of the attribution of intentions with a nonverbal task. *NeuroImage*, **11**, 157–66.

- Cikara, M., Botvinick, M.M., Fiske, S.T. (2011). Us versus them: social identity shapes neural responses to intergroup competition and harm. *Psychological Science*, **22**, 306–13.
- Conway, A.R.A., Conwan, N., Bunting, M.F., Therriault, D.J., Minkoff, S.R.B. (2002). A latent variable analysis of working memory capacity, short-term memory capacity, processing speed, and general fluid intelligence. *Intelligence*, **30**, 163–83.
- Cross, S.E., Bacon, P.L., Morris, M.L. (2000). The relationalinterdependent self-construal and relationships. *Journal of Per*sonality and Social Psychology, **78**(4), 791.
- D'Esposito, M., Postle, B.R., Lease, D.B.J. (1999). Maintenance versus manipulation of information held in working memory: an event-related fMRI study. Brain and Cognition, **41**(1), 66–86.
- Denny, B.T., Kober, H., Wager, T.D., Ochsner, K.N. (2012). A meta-analysis of functional neuroimaging studies of self- and other judgments reveals a spatial gradient for mentalizing in medial prefrontal cortex. *Journal of Cognitive Neuroscience*, 24(8), 1742–52.
- Dodell-Feder, D., Koster-Hale, J., Bedny, M., Saxe, R. (2011). fMRI item analysis in a theory of mind task. *NeuroImage*, **55**(2), 705–11.
- Dufour, N., Redcay, E., Young, L., Mavros, P.L., Moran, J.M., Triantafyllou, C., Saxe, R. (2013). Similar brain activation during false belief tasks in a large sample of adults with and without autism. PlosOne. https://doi.org/10.1371/journal.pone.0075468.
- Eckman, P. (1999). Basic emotions. In: Dalgleish, T., Power, M., editors. Handbook of Cognition and Emotion, Sussex: John Wiley & Sons Ltd, pp. 45–60.
- English, T., Carstensen, L.L. (2014). Selective narrowing of social networks across adulthood is associated with improved emotional experience in daily life. International Journal of Behavioral Development, **38**(2), 1–8.
- Faul, F., Erdfelder, E., Lang, A.-G., Buchner, A. (2007). G*Power 3: a flexible statistical power analysis program for the social, behavioral, and biomedical sciences. *Behavior Research Methods*, **39**, 175–91.
- Fletcher, P.C., Happ´, T., Frith, U., Baker, S.C., Dolan, R.J., Frackowiak, R.S.J., Frith, C.D. (1995). Other minds in the brain: a functional imaging study of theory of mind in story comprehension. Cognition, 57, 109–28.
- Fougnie, D., Marois, R. (2007). Executive working memory load induces intentional blindness. Psychonomic Bulletin & Review, 14(1), 142–7.
- Gallagher, H.L., Happe, F., Brunswick, N., Fletcher, P.C., Frith, U., Frith, C.D. (2000). Reading the mind in cartoons and stories: an fMRI study of 'theory of mind' in verbal and nonverbal tasks. *Neuropsychologia*, **38**(1), 11–21.
- Heleven, E., Overwalle, F. (2019a). Neural representations of others in the medial prefrontal cortex do not depend on our knowledge about them. Social Neuroscience, **14**(3), 286–99.
- Heleven, E., Van Overwalle, F. (2019b). The neural representation of the self in relation to close others using fMRI repetition suppression. Social Neuroscience, https://doi.org/10.3758/s13415-019-00714-2.
- Heleven, E., Van Overwalle, F. (2016). The person within: memory codes for persons and traits using fMRI repetition suppression. Social Cognitive and Affective Neuroscience, **11**(1), 159–71.
- Krienen, F.M., Tu, P.-C., Buckner, R.I. (2010). Clan mentality: evidence that the medial prefrontal cortex responds to close others. The Journal of Neuroscience, 30(41), 13906–15.
- Krol, S.A., Meyer, M.L., Lieberman, M.D., Bartz, J.A. (2018). Social working memory predicts social network size in humans.

Adaptive Human Behavior and Physiology.. doi: 10.1007/s40750-018-0100-9.

- Lewis, P.A., Rezaie, R., Brown, R., Dunbar, R.I.M. (2011). Ventromedial prefrontal volume predicts understanding of others and social network size. *NeuroImage*, **57**(4), 1624–9.
- Maniscalco, B., Lau, H. (2015). Manipulation of working memory contents selectively impairs metacognitive sensitivity in a concurrent visual discrimination task. *Neuroscience of Consciousness.*. doi: 10.1093/nc/niv002.
- Markus, H., Wurf, E. (1986). The dynamic self-concept: a social psychological perspective. Annual Review of Psychology, 38, 299–337.
- McKiernan, K.A., Kaufman, J.N., Kucera-Thompson, J., Binder, J.R. (2003). A parametric manipulation of factors affecting taskinduced deactivation in functional neuroimaging. *Journal of Cognitive Neuroscience*, 15(3), 394–408.
- Meyer, M.L., Spunt, R.P., Berkman, E.T., Taylor, S.E., Lieberman, M.D. (2012). Evidence for social working memory from a parametric functional MRI study. Proceedings of the National Academy of Sciences, 109(6), 1883–8.
- Meyer, M.L., Taylor, S.E., Lieberman, M.D. (2015). Social working memory and its distinctive link to social cognitive ability: an fMRI study. Social Cognitive and Affective Neuroscience, **10**(10), 1338–47.
- Mikels, J.A., Reuter-Lorenz, P.A. (2019). Affective working memory: an integrative psychological construct. Perspectives on Psychological Science, 14(4), 543–59.
- Mitchell, J.P. (2009). Inferences about mental states. Philosophical Transactions of the Royal Society: B, **364**, 1309–16.
- Özdem, C., Brass, M., Schippers, A., Van der Cruyssen, L., Van Overwalle, F. (2019). The neural representation of mental beliefs held by two agents. *Cognitive, Affective, & Behavioral Neuroscience*. https://doi.org/10.3758/s13415-019-00714-2.
- Parkinson, C., Kleinbaum, A., Wheatley, T. (2017). Spontaneous neural encoding of social network position. Nature Human Behavior, 1, 1–7.
- Postle, B.R., Berger, J.S., D'Esposito, M.D. (1999). Functional neuroanatomical double dissociation of mnemonic executive control processes contributing to working memory performance. Proceedings of the National Academy of Sciences, 96(22), 12959–64.
- Postle, B.R., Ferrarelli, F., Hamidi, M., Feredoes, E. (2006). Repetitive transcranial magnetic stimulation dissociates working memory manipulation from retention functions in the prefrontal, but not posterior parietal, cortex. *Journal of Cognitive Neuroscience*, **18**(10), 1712–22.
- Reis, H.T., Shaver, P. (1988). Intimacy as an interpersonal process. In: Duck, S., editor. Handbook of Personal Relationships. Chichester, UK: Wiley, pp. 367–89.
- Roberts, S.G.B., Wilson, R., Fedurek, P., Dunbar, R.I.M. (2008). Individual differences and personal social network size and structure. Personality and Individual Differences, 44(4), 954–64.
- Rohde, T.E., Thompson, L.A. (2007). Predicting academic achievement with cognitive ability. *Intelligence*, **35**(1), 83–92.
- Schurz, M., Radua, J., Aichhorn, M., Richlan, R., Perner, J. (2014). Fractionating theory of mind: a meta-analysis of functional

brain imaging studies. Neuroscience and Biobehavioral Reviews, **42**, 9–34.

- Smith, S.M. (2002). Fast robust automated brain extraction. *Human Brain Mapping*, **17**, 143–55.
- Smith, R., Lane, R., Alkozei, A., Bao, J., Smith, C., Sanova, A., et al. (2017). Maintaining the feelings of others in working memory is associated with activation of the left anterior insula and left frontal-parietal control network. Social Cognitive and Affective Neuroscience.. doi: 10.1093/scan/ nsx011.
- Smith, R., Killgore, W.D.S., Alkozei, A., Lane, R.D. (2018a). A neurocognitive process model of emotional intelligence. Biological Psychology, 139, 131–51.
- Smith, R., Lane, R.D., Sanova, A., Alkozei, A., Smith, C., Killgore, W.D. (2018b). Common and unique neural systems underlying the working memory maintenance of emotional vs. bodily reactions to affective stimuli: the moderating role of trait emotional awareness. Frontiers in Human Neuroscience, 12(370). http://doi:10.3389/fnhum.2018.00370.
- Smith, R., Lane, R., Alkozei, A., Bao, J., Smith, C., Sanova, A., et al. (2018c). The role of medial prefrontal cortex in the working memory maintenance of one's own emotional responses. Scientific Reports, 8, 3460. doi: 10.1038/s41598-018-21896-8.
- Squire, L.R., Zola-Morgan, S. (1991). The medial temporal lobe memory system. *Science*, **253**, 1380–5.
- St. Clair-Thompson, H.L., Gathercole, S.E. (2006). Executive functions and achievements in school: shifting, updating, inhibition, and working memory. *The Quarterly Journal of Experimental Psychology*, **59**(4), 745–59.
- Stiller, J., Dunbar, R.I.M. (2007). Perspective-taking and memory capacity predict social network size. Social Networks, 29(1), 93–104.
- Walter, H., Ciaramidaro, A., Enrici, I., Pia, L., Bara, B.G. (2004). Understanding intentions in social interaction: the role of the anterior paracingulate cortex. *Journal of Cognitive Neuroscience*, 16, 1854–63.
- Waugh, C.E., Lemus, M.G., Gotlib, I.H. (2014). The role of the medial frontal cortex in the maintenance of emotional states. Social Cognitive and Affective Neuroscience, 9(12), 2001–9.
- Wellman, H.M. (1992). The MIT Press series in learning, development, and conceptual change. In: The Child's Theory of Mind, Cambridge: The MIT Press.
- Wimmer, H., Perner, J. (1983). Beliefs about beliefs: representation and constraining function of wrong beliefs in young children's understanding of deception. *Cognition*, 13, 103–28.
- Xin, F., Lei, X. (2015). Competition between frontoparietal control and default networks supports social working memory and empathy. Social Cognitive and Affective Neuroscience, 10(8), 1144–52.
- Yeo, B. T. T., Krienen, F. M., Sepulcre, J., Sabuncu, M. R., Lashkari, D., Hollinshead, M.,... Buckner, R. L. (2011). The organization of the human cerebral cortex estimated by intrinsic functional connectivity. *Journal of Neurophysiology*, **106**, 1125–65. http://dx. doi.org/10.1152/jn.00338.2011.