UNIVERSITY of York

This is a repository copy of *Tracking thoughts::Exploring the neural architecture of mental time travel during mind-wandering*.

White Rose Research Online URL for this paper: <u>https://eprints.whiterose.ac.uk/111403/</u>

Version: Accepted Version

Article:

Karapanagiotidis, Theodoros orcid.org/0000-0002-0813-1019, Bernhardt, Boris C., Jefferies, Elizabeth orcid.org/0000-0002-3826-4330 et al. (1 more author) (2017) Tracking thoughts::Exploring the neural architecture of mental time travel during mind-wandering. Neuroimage. pp. 272-281. ISSN 1053-8119

https://doi.org/10.1016/j.neuroimage.2016.12.031

Reuse

This article is distributed under the terms of the Creative Commons Attribution-NonCommercial-NoDerivs (CC BY-NC-ND) licence. This licence only allows you to download this work and share it with others as long as you credit the authors, but you can't change the article in any way or use it commercially. More information and the full terms of the licence here: https://creativecommons.org/licenses/

Takedown

If you consider content in White Rose Research Online to be in breach of UK law, please notify us by emailing eprints@whiterose.ac.uk including the URL of the record and the reason for the withdrawal request.



eprints@whiterose.ac.uk https://eprints.whiterose.ac.uk/

Tracking thoughts: Exploring the neural architecture of mental time travel during mind-wandering

Theodoros Karapanagiotidis^{a,*}, Boris C. Bernhardt^b, Elizabeth Jefferies^a, Jonathan Smallwood^a

^aDepartment of Psychology and York Neuroimaging Centre, University of York, York, United Kingdom

^bMultimodal Imaging and Connectome Analysis Laboratory, Montreal Neurological Institute and Hospital, McGill University, Montreal, QC Canada

Abstract

The capacity to imagine situations that have already happened or fictitious events that may take place in the future is known as mental time travel (MTT). Studies have shown that MTT is an important aspect of spontaneous thought, yet we lack a clear understanding of how the neurocognitive architecture of the brain constrains this element of human cognition. Previous functional magnetic resonance imaging (MRI) studies have shown that MTT involves the coordination between multiple regions that include mesiotemporal structures such as the hippocampus, as well as prefrontal and parietal regions commonly associated with the default mode network (DMN). The current study used a multimodal neuroimaging approach to identify the structural and functional brain organisation that underlies individual differences in the capacity to spontaneously engage in MTT. Using regionally unconstrained diffusion tractography analysis, we found increased diffusion anisotropy in right lateralised temporo-limbic, corticospinal, inferior fronto-occipital tracts in participants who reported greater MTT. Probabilistic connectivity mapping revealed a significantly higher connection probability of the right hippocampus with these tracts. Resting-state functional MRI connectivity analysis using the right hippocampus as a seed region revealed greater functional coupling to the anterior regions of the DMN with increasing levels of MTT. These findings demonstrate that the interactions between

*Corresponding author Email address: tk710@york.ac.uk (Theodoros Karapanagiotidis)

Preprint submitted to Neuroimage

the hippocampus and regions of the cortex underlie the capacity to engage in MTT, and support contemporary theoretical accounts that suggest that the integration of the hippocampus with the DMN provides the neurocognitive landscape that allows us to imagine distant times and places.

Keywords: Mental time travel, Spontaneous thought, Diffusion MRI, Hippocampus, Mesiotemporal lobe, Default mode network

1 1. Introduction

Conscious experience is not restricted to events in the here and now. 2 The prominence of states such as daydreaming and mind-wandering in our 3 mental lives illustrates that we often mentally escape from the constraints of 4 the moment and generate thoughts regarding people, places and situations other than those in the immediate environment (Killingsworth and Gilbert, 6 2010). It is now known that these states are accompanied by a measurable reduction in the processing of external events, indicating a disengagement, 8 or decoupling, of attention from the external environment (Smallwood et al., 9 2008). Instead of relying on perceptual input for their mental content, it is 10 hypothesised that these experiences are built almost exclusively from repre-11 sentations in semantic and episodic memory (see Smallwood and Schooler 12 2015 for a review). 13

Evidence from functional neuroimaging is consistent with the view that 14 memory retrieval is a core process with which we generate spontaneous 15 thought. Prior work has demonstrated that a large-scale brain system known 16 as the default mode network (DMN) is important for the thoughts that are 17 generated during the mind-wandering state (Mason et al., 2007; Christoff 18 et al., 2009). The DMN is generally identified as a distributed regional 19 assembly anchored by hubs in the medial prefrontal cortex, the posterior 20 cingulate cortex and the angular gyrus (Raichle et al., 2001; Raichle, 2015). 21 During tasks that involve retrieving information from memory, the DMN of-22 ten integrates information from medial and lateral temporal regions known to 23 play a pivotal role in episodic and semantic memory, such as the hippocam-24 pus (Sestieri et al., 2011) and the anterior temporal lobe (Patterson et al., 25 2007). Several studies have linked the DMN, as well as related structures 26 such as the hippocampus, to processes such as imagining events from the 27 future or past (Schacter et al., 2007), which are collectively known as mental 28 time travel (MTT). Studies have shown that MTT is an important element 29

of the mind-wandering state (Smallwood et al., 2009b; Baird et al., 2011)
and a meta analytic study has shown similarities between the neural activation during mind-wandering and episodic future thinking (Stawarczyk and
D'Argembeau, 2015). A recent study (Ellamil et al., 2016) has also indicated
that the hippocampus is activated early during the spontaneous generation
of thoughts while mind-wandering.

Although previous functional magnetic resonance imaging (MRI) studies 36 have established a functional role of the DMN in the mind-wandering state 37 (Mason et al., 2007; Allen et al., 2013; Ellamil et al., 2016), the extent to 38 which the neural architecture of this network and associated regions of cor-39 tex constrain naturally occurring spontaneous thought remains uncertain. 40 In the current study, we tested the hypothesis that variations in sponta-41 neous thoughts across participants emerge as a consequence of the large-scale 42 structural network organisation of the brain. We applied regionally uncon-43 strained tractography analysis to diffusion magnetic resonance imaging data 44 in a cohort of healthy adults and assessed whether individual variability in 45 the contents of their thoughts related to markers of structural connectiv-46 ity. Our analysis approach was complemented with probabilistic connectiv-47 ity mapping, to identify the cortical grey matter with the highest connection 48 probability to these tracts. Finally, we used seed-based resting-state func-49 tional MRI connectivity mapping to identify associated functional networks. 50 Based on functional studies of the role of the DMN and regions in the medial 51 and lateral temporal lobe in thoughts generated during the mind-wandering 52 state (Christoff et al., 2009; Stawarczyk et al., 2011), we anticipated our 53 structural analysis to highlight a constellation of regions, including the me-54 dial prefrontal, posterior cingulate, as well as lateral and medial temporal 55 cortices, including the hippocampus. 56

57 2. Methods

58 2.1. Participants

⁵⁹ A total of 86 healthy participants were recruited by advert from the De-⁶⁰ partment of Psychology at the University of York (51 women, age range 18 -⁶¹ 31). They were offered either a payment of £20 or a commensurate amount ⁶² of course credits. Written consent was obtained for all participants and the ⁶³ study was approved by the York Neuroimaging Centre Ethics Committee. ⁶⁴ Participants were recruited in two cohorts (Sample A, n = 47; Sample B, n ⁶⁵ = 39) in different time periods, although there were no differences relevant

to the study. While both samples participated in the behavioural session 66 and underwent a resting state (rs) functional MRI scan, we acquired diffu-67 sion MRI data only for sample A. We excluded 8 out of 39 participants from 68 Sample B due to incomplete brain coverage (whole-brain coverage < 94%). 69 Having two samples gave us the opportunity to treat them separately in our 70 analyses and investigate the robustness of our behavioural and functional 71 MRI results. The behavioural and functional MRI data in this study are the 72 same as those reported in Smallwood et al. (2016). 73

74 2.1.1. Independent sample

We also used an independent dataset to provide independent confirmation of functional connectivity results. These data were obtained from a publicly available dataset: the Nathan Kline Institute (NKI)/Rockland Enhanced Sample and contained 141 subjects. Full details of this sample can be found in Gorgolewski et al. (2014).

80 2.2. Behavioural Methods

81 2.2.1. Choice Reaction Time Task

To acquire information about the content of spontaneous thought in a 82 situation conducive to the mind-wandering state, participants performed a 83 simple non-demanding choice reaction time task. This task is routinely used 84 in studies of spontaneous thought because it creates periods when sponta-85 neous thoughts are generated with as similar a frequency as when participants 86 are not engaging in a task (Smallwood et al., 2009b). Participants sat in a 87 testing booth and were asked to make a parity judgement to numerals that 88 were coloured red. These stimuli were presented in a stream of non-coloured 89 numerals, to which no response was required. Stimuli were presented with a 90 slow inter-stimulus interval (2200-4400 ms) and remained on screen for 1000 91 ms. The task lasted 15 min and participants performed a single run. The 92 occurrence of the target and non-target stimuli was randomly determined 93 with a mean target number of M = 25.2, SE = 0.6 and a mean non-target 94 number of M = 109.6, SE = 1.1. Participants responded by using the mouse 95 button. Accuracy was high (mean \pm SD = 0.93 \pm 0.08), with a mean response 96 time of 900 ms (SD = 161). Participants performed this task in a laboratory 97 testing session, scheduled one day after the scanning. 98

⁹⁹ 2.2.2. Multi-Dimensional Experience Sampling (MDES)

At unpredictable moments while performing the laboratory task, partici-100 pants were interrupted and asked to rate different aspects of their experience. 101 They were asked to focus their answers on the contents of their experience 102 the moment immediately prior to the interruption, thereby reducing demands 103 on memory. They responded using a continuous Likert scale. The specific 104 questions used are described in Table 1 and they were all asked during each 105 probing. They were selected from prior studies (Smallwood et al., 2016) and 106 examined the content of thoughts (e.g. temporal content relating to the past 107 or future, referent of thought - themselves or another person - and emotional 108 valence), as well as the form these thoughts took (whether the thoughts were 109 in words or images, the level of detail and intrusiveness, etc.). 110

Whenever experience sampling occurred, the questions were administered 111 in a quasi-random order. The first question was always about task focus, 112 followed by blocks of questions about the content and form of thoughts. 113 On each occasion, the order of each block, as well as the order of questions 114 within each block, was randomised. Participants were probed an average of 8 115 times during the fifteen-minute task. We used a fully randomised sequence of 116 experience sampling probes to ensure that regularities in our probing schedule 117 did not bias the results of our experiment (Seli et al., 2013). 118

As in previous studies, data from each individual was concatenated into 119 a single matrix. We z-scored the data from each sample separately to min-120 imise differences between them, and fed them into a principal component 121 analysis (PCA) with varimax rotation, allowing patterns of covariance that 122 broadly correspond to different types of thought to be identified. The decom-123 position of these data, as well as a replication sample, has previously been 124 reported by Smallwood et al. (2016). The decomposition was performed in 125 separate analyses, one for the content of the experience and one for the form. 126 We chose to decompose the components of content and form separately be-127 cause our method of experience sampling presented them in these conceptual 128 groups, reflecting our a priori interest in decomposing them as separate fac-129 tors. In addition, we have successfully employed this technique of experience 130 sampling in several prior studies (Ruby et al., 2013a,b; Engert et al., 2014; 131 Medea et al., 2016) and applying the same procedure in the current study 132 provided the chance to relate our data to these prior investigations. Table 133 S1 in the supplementary materials presents the eigenvalues for the first four 134 components generated from the decomposition of the content and the form 135

questions, where it can be seen that the scores change rapidly after the third
component. For this reason, and to remain consistent with these previous
studies, we opted for a three-component solution.

139 2.3. Neuroimaging Methods

140 2.3.1. MRI data acquisition

MRI data were acquired on a GE 3 Tesla Signa Excite HDxMRI scanner, 141 equipped with an eight-channel phased array head coil at York Neuroimaging 142 Centre, University of York. For each participant, we acquired a sagittal 143 isotropic 3D fast spoiled gradient-recalled echo T1-weighted scan (TR = 7.8144 ms, TE = minimum full, flip angle = 20° , matrix = 256x256, voxel size = 145 $1.13 \times 1.13 \times 1 \text{ mm}^3$, FOV = $289 \times 289 \text{ mm}^2$). Resting-state functional MRI data 146 based on blood oxygen level-dependent contrast images with fat saturation 147 were acquired using a gradient single-shot echo-planar imaging sequence (TE 148 = minimum full (≈ 19 ms), flip angle = 90°, matrix = 64x64, FOV = 192x192 149 mm^2 , voxel size = $3x3x3 mm^3$). Sample A had a scan duration of 9 min 150 and the following additional parameters TR = 3000 ms, 180 volumes, slice 151 thickness 3 mm, no gap, 60 slices. Sample B had a scan duration of 7 min, 152 TR = 2000 ms, 210 volumes, slice thickness 3 mm, 0.5 mm gap and 32153 slices. The duration of the diffusion MRI scan was 13 minutes. A single-shot 154 pulsed gradient spin-echo echo-planar imaging sequence was used with the 155 following parameters: $b = 1000 \text{ s/mm}^2$, 45 directions, 7 T2-weighted EPI 156 baseline scans (b0), 59 slices, $FOV = 192 \times 192 \text{ mm}^2$, TR = 15 s, TE = 86 ms157 (minimum full), voxel size = $2x2x2 \text{ mm}^3$, matrix = 96x96. 158

159 2.3.2. Structural connectivity analysis

Diffusion MRI data pre-processing involved eddy-current distortion cor-160 rection and motion correction using FDT v3.0, part of FSL (Smith et al., 161 The fractional anisotropy (FA) was calculated by fitting a tensor 2004). 162 model at each voxel of the pre-processed diffusion data and the resulting 163 images were brain-extracted using BET (Smith, 2002). Voxel-wise FA maps 164 were analysed using Tract-Based Spatial Statistics (TBSS) (Smith et al., 165 2006). No advanced options were used. After subjects' FA data were non-166 linearly aligned to the FMRIB58 template in MNI152 space, the mean FA im-167 age was created and thinned to create a mean FA skeleton, which represents 168 the centres of all tracts common to the group. Using a generalised model, 169 we assessed correlations between measured FA values across the skeleton and 170 the mind-wandering PCA scores of each participant. T-statistics maps for 171

contrasts of interest were calculated using FSL's Randomize (a nonparametric permutation inference tool) with 5000 permutations (Nichols and Holmes, 2002). Resulting maps were thresholded at a Family-Wise Error (FWE) corrected p < 0.05 using Threshold-Free Cluster Enhancement (TFCE) (Smith and Nichols, 2009).

We fitted voxel-wise probabilistic diffusion models using BEDPOSTX 177 (Behrens et al., 2003) with 2 fibres modelled per voxel and 1000 itera-178 tions. Subsequent to BEDPOSTX, probabilistic tractography was performed 179 using PROBTRACKX (Behrens et al., 2007) to reconstruct fibres pass-180 ing through a single-mask or connecting two masks at a time. Tractogra-181 phy was performed in native diffusion space. To this end, we transformed 182 our seed masks from standard space back to diffusion space using the in-183 verse of the nonlinear registration calculated in the TBSS pipeline. PROB-184 TRACKX was used with standard parameters (5000 samples/voxel, cur-185 vature threshold 0.2, step length 0.5 mm, samples terminated after 2000 186 steps or when they reached the surface as defined by a 40% probabilistic 187 whole-brain WM mask). In the single-mask case, the connectivity maps of 188 each individual were thresholded at 1% of total samples sent from the seed 189 mask, mapped back to standard space using nonlinear registration, and con-190 catenated into a single 4D file. Nonparametric voxelwise statistical testing 191 with 25000 permutations was then performed to obtain a group-level prob-192 abilistic tractography map, thresholded using TFCE at p < 0.05, FWE-193 corrected as above. In the dual-mask case, we performed seed-to-target 194 analyses, with atlas volumes as the seeds and clusters of significant find-195 ings from our analyses as the targets. We also ran seed-to-target analyses 196 using diffusion imaging data (b = 1500 s/mm^2 , 127 directions plus 9 in-197 terspersed b0 images, voxel size = 2x2x2 mm³) from a subset (n = 9, 4 198 women, age range 21 - 48) of the Test-Retest Pilot Dataset/enhanced NKI 199 sample. Full details of this sample, as well as all the parameters of the 200 diffusion-weighted imaging sequence used, can be found online here: http: 201 //fcon 1000.projects.nitrc.org/indi/pro/eNKI RS TRT/FrontPage.html. 202

203 2.3.3. Functional connectivity analysis

Functional MRI pre-processing and analyses were performed using FSL. Following the co-registration of functional and structural data, we extracted the brain using BET and linearly registered them to MNI152 space. Prior to functional connectivity analysis, resting state data underwent motion correction using MCFLIRT, slice-timing correction using Fourier-space time-series phase shifting, brain extraction using BET, spatial smoothing using a Gaussian kernel with a full width at half maximum of 6 mm, grand-mean intensity normalisation followed by high-pass (sigma = 100 s) and low-pass temporal filtering (sigma = 2.8 s).

In each subject, we extracted the time series from seed regions of interest (i.e. atlas volumes or significant clusters identified in previous steps) and used them as explanatory variables in separate functional connectivity analyses that also included 11 nuisance regressors: the top five principal components extracted from WM and cerebrospinal fluid masks in accordance with the CompCor method (Behzadi et al., 2007) and six motion parameters. No global signal regression was performed (Murphy et al., 2009).

Group-level statistical modelling was carried out using FEAT/FLAME 220 stage 1 (Woolrich et al., 2004) with automatic outlier detection (Woolrich, 221 2008). A 50% probabilistic GM mask was applied and results were thresh-222 olded at the whole-brain level using cluster-based Gaussian random field 223 theory, with a cluster-forming threshold of z > 3.1 (and z > 2.3 when 224 the two samples were analysed independently) and a FWE corrected clus-225 ter significance level of p < 0.05. To further confirm our group-level find-226 ings we also ran the analysis using permutation testing with the obtained 227 maps thresholded using TFCE at p < 0.05, FWE-corrected. Unthresh-228 olded maps were uploaded onto Neurovault and can be found here: http: 229 //neurovault.org/collections/1448. 230

231 3. Results

232 3.1. Analysis aims

The goal of this experiment was to identify connections between the struc-233 tural and functional organisation of the brain and variations in different types 234 of spontaneous thought. We first calculated the principal components of 235 the type of spontaneous thoughts as reported in a laboratory session. Next 236 we determined whether there was any relationship between the fractional 237 anisotropy of tractography-derived white matter tracts and inter-individual 238 variation in the content and form of spontaneous thought. Finally, we ex-230 plored the functional connectivity of grey matter regions that received pro-240 jections from the tracts identified in the previous step, with the aim of iden-241 tifying whether the functional connectivity of these regions is modulated by 242 the same aspects of spontaneous thought. The analysis pipeline is outlined 243 in Figure 1. 244

245 3.2. Components of spontaneous thought

For the decomposition of content, we focused on questions relating to 246 temporal focus, referent of thought, task focus, and emotional content. Con-247 sistent with prior investigations (Ruby et al., 2013a,b; Engert et al., 2014; 248 Medea et al., 2016), we found three orthogonal factors: i) Future and self-249 focused thoughts: individuals with high weighting on this component were 250 often thinking about themselves in the future, accounting for 29% of the 251 observed variance; ii) Past-focused social thoughts: individuals with high 252 weighting were often thinking about self and others in the past, accounting 253 for 19% variance; iii) Task-related thoughts: individuals with high weighting 254 were often thinking about the task itself and experienced fewer negatively 255 valenced off-task thoughts, accounting for 18% variance. The average of the 256 future and past components, which we refer to as MTT, accounted for 48%257 of the overall variance. 258

Our next step was to decompose the questions regarding the form of 259 thoughts - such as whether these were experienced as images or words, if they 260 were detailed and whether they were intrusive - following a similar procedure. 261 This yielded three components: i) The modality of the thoughts (images or 262 words): individuals with high weighting often described their thoughts as 263 containing words rather than images and this reflected 33% of the variance; 264 ii) The level of intrusiveness of the thoughts: individuals with high weighting 265 often described their thoughts as intrusive, accounting for 26%; iii) The level 266 of detail in the thoughts: individuals with low weighting on this reported 267 more detail in their thoughts accounting for 23%. These patterns of the form 268 of cognition are consistent with prior investigations (Medea et al., 2016; see 269 also the replication sample in Smallwood et al., 2016). 270

271 3.3. Identifying the relationship between white matter fractional anisotropy 272 and the contents of spontaneous thought

Relating component weighting of the mind-wandering PCA scores to 273 skeleton-wide FA values derived from the TBSS analysis revealed a spe-274 cific increase in the fractional anisotropy of a temporo-limbic white mat-275 ter region for people engaging more in MTT (Figure 2i-top, Table 2). No 276 other components showed any significant association. In order to describe 277 the whole-brain structural connectivity profile of this region, we performed 278 a probabilistic tractography analysis using it as a seed (Figure 2i-bottom). 279 This analysis showed (Figure 2ii) that the clusters' structural connectogram 280 closely overlapped with the right fornix, the right corticospinal tract, and 281

the right inferior fronto-occipital fasciculus, as defined by the Johns Hopkins University DTI-based white-matter atlases (Mori et al., 2005).

As this region falls in an area with a high degree of crossing fibres, we 284 also performed an additional analysis using a model that incorporates fibre-285 specific measurements (tbss x) (Jbabdi et al., 2010). We estimated the 286 primary and secondary fibre orientations, together with their partial volume 287 fractions, and found a significant increase in the partial volume fraction of 288 the primary orientation for people engaging more in MTT. No other compo-289 nent was significant for any of the two orientations. The identified regions, 290 presented in the supplementary materials, include the areas discovered in our 291 original analysis, but being more widespread did not improve their structural 292 classification. For this reason and due to the potential limitations of our dif-293 fusion imaging sequence in regard to probabilistic analyses and crossing fibres 294 (see Discussion), we did not consider them any further. 295

To identify those grey matter regions most likely to be connected to the 296 temporo-limbic white matter substrate of MTT, we used the seeds-to-target 297 mode of PROBTRACKX with no advanced options and calculated proba-298 bilistic streamline counts that reach our target mask when seeding from 116 299 regions-of-interest, as defined by the automated anatomical labelling (AAL) 300 (Tzourio-Mazoyer et al., 2002) for a given participant. This generated a 116 301 x 47 connectivity matrix, in which streamline counts were normalised by 302 the total number of generated tracts (waytotal), thus translated to connec-303 tion probabilities. The results demonstrate that the most likely grey matter 304 region connected with the MTT substrate is the right hippocampus (Fig-305 ure 3). To quantitatively assess this, we calculated the difference between 306 the hippocampus connection probability to our target mask and the second 307 highest connection probability, per participant. The generated distribution 308 was significantly greater than 0 (one-sample Wilcoxon Signed Ranked test, 309 p < 0.0001). Arguably, the TBSS-derived cluster was situated in a white-310 matter region that may harbour extensive fibre crossing, which may challenge 311 tractography-based fibre reconstruction, particularly when a low number of 312 diffusion directions is used. We therefore repeated the diffusion tractography 313 analyses based on data from the NKI Enhanced repository (see Structural 314 connectivity analysis), which were acquired using a sequence with a substan-315 tially increased number of diffusion directions. Theoretically, such sequences 316 should better resolve crossing-fibres and thus minimise the risk of erroneous 317 tractography results (see Discussion). Our analyses at this higher angular 318 resolution confirmed that the right hippocampus was the grey matter region 319

most likely to be connected to the temporo-limbic white matter substrate of MTT, supporting our initial findings.

322 3.4. Determining the link between the functional connectivity of the hip-323 pocampus and the content of spontaneous thought

Having identified the hippocampus as the region with the highest connec-324 tion probability to our white matter MTT substrate, we assessed its func-325 tional connectivity profile and the modulation of these patterns by inter-326 individual differences in the propensity to engage in MTT. Our motivation 327 for performing this analysis was two-fold. First, we wanted to test the ro-328 bustness of the diffusion imaging results by investigating whether the selected 329 region had a specific relationship to MTT using data from another scanning 330 modality. Second, we wished to understand whether the mechanism that un-331 derlies the role of the hippocampus in MTT was related to its integration into 332 the DMN. To assess these two questions we took advantage of the fact that 333 we had two cohorts of participants for whom resting state functional MRI 334 data were available and who also had MDES descriptions of their thoughts. 335 We calculated the functional connectivity of the AAL mask of the right hip-336 pocampus for each participant in each cohort. These maps were used as the 337 dependent variables in a multiple regression analysis with the 6 dimensions 338 from the PCA decomposition of MDES scores used as independent variables, 339 all in the same model. 340

Group-level functional connectivity of the right hippocampus indicated 341 extensive connections to the ventromedial prefrontal cortex, the retrosplenial 342 cortex, the brain stem and the cerebellum (Figure 4i). Next, we assessed re-343 gions whose connectivity with the right hippocampus correlated with individ-344 ual differences in the content and form of participants' thoughts. This anal-345 ysis revealed a region of dorsal anterior cingulate cortex/medial pre-frontal 346 cortex as can be seen in Figure 4i (cluster: size 646 voxels, volume 5168 347 mm³, centre of gravity -4, 48, 11 mm), whose connectivity to the hippocam-348 pus increased for individuals with higher MTT scores (MTT+) compared 349 to those with lower scores. These results were thresholded at the whole-350 brain level with a cluster-forming threshold of z > 3.1 and a FWE corrected 351 cluster significance level of p < 0.05. The same regions (Figure S3 in supple-352 mentary materials) were also highlighted for the MTT+ comparison from our 353 group-level permutation testing analysis. Finally, we observed two significant 354 clusters of functional connectivity that were associated with the modality of 355

thoughts and their relative levels of detail. These did not survive more stringent analyses, so we do not consider them further, however we present their spatial maps in the supplementary materials, and the unthresholded maps can be found in Neurovault.

We also repeated these analyses separately in both datasets to investi-360 gate whether idiosyncratic features of one sample may have contributed to 361 our findings. Comparing each group separately, only Sample B passed a clus-362 ter forming threshold of z > 3.1, however both groups produced significant 363 cluster corrected regions at a more lenient value of z > 2.3. In both cases 364 these regions fell in the mPFC and are shown in Figure 4ii, alongside their 365 overlap (cluster: size 82 voxels, volume 656 mm³, centre of gravity -4, 47, 366 11 mm) and separate scatter plots from each sample. These analyses show 367 a robust pattern of strengthened correlation between the right hippocampus 368 and the medial prefrontal cortex for individuals who engage in greater MTT 360 when the sample is treated as a group; however, we also find a converging 370 pattern when both groups are treated as separate populations. These two 371 features of our data suggest that our findings are statistically robust and are 372 reasonably consistent across both groups of participants. 373

The observation that coupling between the right hippocampus and the 374 medial prefrontal cortex was greater for participants who engaged in increas-375 ing levels of MTT suggests that this experience may depend upon integration 376 between the medial temporal lobe and the DMN. To quantitatively assess 377 this possibility, we used the overlap region identified in the previous step as 378 a seed in a functional connectivity analysis of an independent dataset (see 379 2.1.1 "Independent sample"). This analysis confirmed a pattern of connec-380 tivity focused on the posterior cingulate and medial prefrontal cortex, which 381 reflects the canonical DMN (Figure 4iii). 382

383 4. Discussion

We demonstrated that the content and form of spontaneous thought is 384 partly constrained by structural and functional brain network organisation. 385 Structural connectivity analyses identified a temporo-limbic white matter 386 region, highly connected to the right hippocampus, for people who spon-387 taneously engaged in more mental time travel (MTT). Using resting state 388 functional connectivity, we found that the temporal correlation of the right 389 hippocampus with the dorsal anterior cingulate cortex, a core region of the 390 default mode network (DMN), was also modulated by inter-individual vari-391

ation in MTT. These converging lines of evidence provide unambiguous support that the spontaneous thoughts experienced during the mind-wandering
state are in fact reliant upon the hippocampus and its integration into the
DMN.

Component process accounts of the mind-wandering state suggest that 396 this class of experience depends upon distinct neurocognitive components. 397 These include the ability to disengage attention from external input, known 398 as perceptual decoupling, and processes more directly related to the genera-399 tion and coordination of the experiential content (Smallwood and Schooler, 400 2015). Among them, episodic memory processes have been considered to play 401 a key role in the generation of the mental content during mind-wandering, 402 particularly those episodes that entail imagining distant times and places 403 (MTT). People frequently use MTT to consider autobiographical goals (Baird 404 et al., 2011), solve personal problems (Ruby et al., 2013b), reduce social stress 405 (Engert et al., 2014) and to generate creative solutions to problems (Baird 406 et al., 2012): all tasks that draw on multiple types of memory and involve the 407 hippocampus. The hippocampus has been linked to a broad range of cogni-408 tive processes, including episodic or autobiographical memory (Eichenbaum, 409 1993; Aggleton and Brown, 1999), spatial navigation (O'keefe and Nadel, 410 1978; O'Keefe et al., 1996) and the binding of temporally extended events 411 into a sequence (for a review see Eichenbaum 2013). These distinct accounts 412 have led to the proposal that the hippocampus may serve an integrative 413 function in cognition by combining information from different domains to 414 form coherent scenes (Hassabis and Maguire, 2007; Maguire et al., 2015), 415 allowing autobiographical information to be placed in a temporal and spatial 416 context (Eichenbaum and Cohen, 2014). Building on this view, it is possible 417 that the hippocampus integrates different aspects of knowledge from mem-418 ory into an ongoing train of thought. It could do so by its dense structural 419 and functional connectivity profile to multiple areas of cortex (Squire et al., 420 2004; Moscovitch et al., 2016; Bernhardt et al., 2016; Strange et al., 2014). 421 Our demonstration that the structural and functional connectivity of the 422 hippocampus is important in MTT may reflect a hippocampal contribution 423 to the process through which we use our memory to consciously organise our 424 life goals and evaluate our past experiences. It is worth noting that, although 425 our main diffusion tractography analysis targeted the right hippocampus, our 426 supplementary analysis after accounting for crossing fibres highlighted tracts 427 more symmetrical across hemispheres (see supplementary Figure S1), sug-428 gesting an involvement of both left and right hippocampi. Thus, while prior 429

studies have found that the right hippocampus is important in mental time
travel especially when imagining events (see Arzy et al. 2009, Experiment
2, see also Addis and Schacter 2012 for further consideration of this issue),
further work is needed to identify the role of the hippocampus in different
hemispheres in spontaneous mental time travel.

Our functional data showed that the region with heightened hippocam-435 pal coupling for increased mental time travel fell outside of the group con-436 nectivity map of the hippocampus (see Figure 4). This suggests that the 437 contribution of the hippocampus to MTT involves integration with the me-438 dial prefrontal cortex, a core node of the DMN (Gusnard and Raichle, 2001; 439 Greicius et al., 2003; Buckner et al., 2008). Contemporary accounts of this 440 network (Andrews-Hanna et al., 2014a) propose that the DMN consists of 441 discrete subsystems, whose coupling to the medial core - the medial pre-442 frontal cortex and posterior cingulate cortex - influences ongoing cognitive 443 processing (Andrews-Hanna et al., 2014b). In concordance with this view, 444 it has been shown that, during memory retrieval, the hippocampus couples 445 with other DMN regions more strongly than it does at rest (Huijbers et al., 446 2011), while hippocampal - DMN interactions have also been highlighted in 447 conceptual processing (Constantinescu et al., 2016). Our demonstration that 448 increased functional connectivity between the hippocampus and the medial 449 prefrontal cortex is associated with increased MTT supports this component 450 process view of the DMN function, suggesting that engaging in spontaneous 451 episodic thought is a situation when the hippocampus and the DMN act to-452 gether in an integrated fashion. Our data from the domain of spontaneous 453 thought provides further evidence that the DMN is important in attending 454 to distant times and places (Peer et al., 2015), perhaps because it functions 455 to integrate information from across the cortex (Margulies et al., 2016). This 456 hypothesis could be further explored by looking at the content of thought 457 while ongoing measures of neural function are recorded (Tusche et al., 2014). 458 In addition, as recent studies have demonstrated ways of measuring activity 459 in white matter from a functional MRI acquisition (Gawryluk et al., 2014; 460 Ding et al., 2016), it could be of interest to explore the temporal correla-461 tions along white matter tracts and how these might relate to spontaneous 462 thoughts and DMN connectivity. 463

There are some limitations that should be borne in mind when considering our data. First, we only measured the functional and structural organisation of neural functioning in the participants on one occasion. Although the converging evidence produced by two independent imaging methodologies, and

especially the diffusion data, show that these data are most parsimoniously 468 described as reflecting a trait, there remains a possibility that the experience 469 sampling observations are partly influenced by state related changes. Fu-470 ture studies may gain greater power by measuring experience across several 471 days, which would provide measures of the content of mind-wandering that 472 are more closely tied to an individual's trait. Also, our study explored the 473 neural correlates of latent patterns within a multi dimensional experience 474 sampling space by performing a data reduction using principal components 475 analysis. This allows us to characterise the largest patterns within the expe-476 rience sampling data in a statistically robust manner; however, its weakness 477 is that it does not provide the specificity to determine whether it is the self-478 relevant or temporal aspects of cognition, or a combination of both, that our 479 data capture. Future work with a larger sample size could profitably explore 480 this issue by modelling the interactions at the level of each question. In addi-481 tion, our whole-brain tractographic findings did not survive a superordinate 482 Bonferonni correction additionally adjusting family-wise error levels for the 483 number of different contrasts included in our model. We believe that this is 484 quite possibly due to the sample size of our diffusion MRI data and unlikely 485 to be a Type 1 error, as the MTT score was found to be a significant predictor 486 of a fractional anisotropy increase in a white matter region highly connected 487 to the hippocampus, a region for which there are strong a priori reasons to 488 expect it to play an important role in the mind-wandering state (Schacter 489 et al., 2007). Moreover, the functional connectivity of the hippocampus to 490 core regions of the DMN showed a similar pattern of modulation by MTT 491 across two datasets. Finally, it is also important to note that the current 492 study was carried out using a conventional diffusion imaging sequence with 493 45 diffusion directions and only one b-value shell. While it has been argued 494 previously (Jones, 2004; Jones et al., 2013) that 30 or more unique orienta-495 tions allow to obtain robust estimates of tensor-derived properties (fractional 496 anisotropy and principal eigenvector orientation), tractographic analysis and 497 estimation of tensor parameters may generally be challenged when different 498 fibre populations cross in a given voxel. These limitations motivate more 499 targeted follow-up diffusion MRI studies on the observed relation, that can 500 take advantage of increased angular resolution by moving to higher fields, us-501 ing longer scans, and/or by utilizing accelerated image acquisition techniques 502 (Feinberg et al., 2010). 503

In conclusion, our study highlights that although spontaneous thoughts seem to emerge independently of external input, they are nonetheless con-

strained by the structure of the cortex. Our results provide evidence that 506 the connections between the hippocampus and other areas of the cortex sup-507 port the contribution of episodic content during spontaneous thought. These 508 findings complement prior studies linking individual differences in sponta-509 neous thought to (i) neural measures such as cortical thickness (Bernhardt 510 et al., 2014), functional connectivity (Smallwood et al., 2013a; Tusche et al., 511 2014; Smallwood et al., 2016), and (ii) psychological measures such as ex-512 ecutive control (Smallwood et al., 2013b; Kane et al., 2007; Levinson et al., 513 2012; McVay and Kane, 2009) and personality (McVay et al., 2009; Diaz 514 et al., 2014; Golchert et al., 2016). Together, these complimentary lines of 515 research demonstrate that although the specific content our thoughts take is 516 doubtlessly influenced by our current concerns (Klinger and Cox, 1987) or 517 our mood (Smallwood et al., 2009a; Killingsworth and Gilbert, 2010; Poerio 518 et al., 2013), aspects of how they emerge may be determined by more stable 510 aspects of a person. Thus, even though our clear sense is that spontaneous 520 thoughts emerge from nowhere (Schooler, 2002), the manner in which this 521 process occurs is likely influenced by the organisation of our neurocognitive 522 system. 523

524 Acknowledgements

TK was supported by a doctoral studentship of the department of Psy-525 chology of the University of York. BB was supported by the Montreal Neu-526 rological Institute and the Canadian Institutes of Health Research. EJ was 527 supported by grants from BBSRC (BB/J006963/1) and the European Re-528 search Council (SEMBIND - 283530) and JS was supported by European 529 Research Council (WANDERINGMINDS - 646927). This publication was 530 also made possible through the support of a grant from the John Templeton 531 Foundation, "Prospective Psychology Stage 2: A Research Competition" to 532 Martin Seligman. The opinions expressed in this publication are those of 533 the author(s) and do not necessarily reflect the views of the John Templeton 534 Foundation. 535

536 References

Addis, D. R., Schacter, D., 2012. The hippocampus and imagining the future:
where do we stand? Frontiers in human neuroscience 5, 173.

Aggleton, J. P., Brown, M. W., 1999. Episodic memory, amnesia, and the
hippocampal-anterior thalamic axis. Behavioral and brain sciences 22 (03),
425-444.

Allen, M., Smallwood, J., Christensen, J., Gramm, D., Rasmussen, B.,
Gaden Jensen, C., Roepstorff, A., Lutz, A., 2013. The balanced mind: the
variability of task-unrelated thoughts predicts error-monitoring. Frontiers
in human neuroscience 7, 743.

Andrews-Hanna, J. R., Saxe, R., Yarkoni, T., 2014a. Contributions of
episodic retrieval and mentalizing to autobiographical thought: evidence
from functional neuroimaging, resting-state connectivity, and fmri metaanalyses. Neuroimage 91, 324–335.

Andrews-Hanna, J. R., Smallwood, J., Spreng, R. N., 2014b. The default
network and self-generated thought: component processes, dynamic control, and clinical relevance. Annals of the New York Academy of Sciences
1316 (1), 29–52.

Arzy, S., Collette, S., Ionta, S., Fornari, E., Blanke, O., 2009. Subjective
mental time: the functional architecture of projecting the self to past and
future. European Journal of Neuroscience 30 (10).

Baird, B., Smallwood, J., Mrazek, M. D., Kam, J. W., Franklin, M. S.,
Schooler, J. W., 2012. Inspired by distraction mind wandering facilitates
creative incubation. Psychological Science, 0956797612446024.

Baird, B., Smallwood, J., Schooler, J. W., 2011. Back to the future: autobiographical planning and the functionality of mind-wandering. Consciousness
and cognition 20 (4), 1604–1611.

Behrens, T., Berg, H. J., Jbabdi, S., Rushworth, M., Woolrich, M., 2007.
Probabilistic diffusion tractography with multiple fibre orientations: What
can we gain? Neuroimage 34 (1), 144–155.

Behrens, T., Woolrich, M., Jenkinson, M., Johansen-Berg, H., Nunes, R.,
Clare, S., Matthews, P., Brady, J., Smith, S., 2003. Characterization and
propagation of uncertainty in diffusion-weighted mr imaging. Magnetic resonance in medicine 50 (5), 1077–1088.

Behzadi, Y., Restom, K., Liau, J., Liu, T. T., 2007. A component based noise
correction method (compcor) for bold and perfusion based fmri. Neuroimage 37 (1), 90–101.

Bernhardt, B. C., Bernasconi, A., Liu, M., Hong, S.-J., Caldairou, B.,
Goubran, M., Guiot, M. C., Hall, J., Bernasconi, N., 2016. The spectrum of structural and functional imaging abnormalities in temporal lobe
epilepsy. Annals of neurology.

Bernhardt, B. C., Smallwood, J., Tusche, A., Ruby, F. J., Engen, H. G.,
Steinbeis, N., Singer, T., 2014. Medial prefrontal and anterior cingulate
cortical thickness predicts shared individual differences in self-generated
thought and temporal discounting. Neuroimage 90, 290–297.

Buckner, R. L., Andrews-Hanna, J. R., Schacter, D. L., 2008. The brain's default network. Annals of the New York Academy of Sciences 1124 (1), 1–38.

⁵⁸⁴ Christoff, K., Gordon, A. M., Smallwood, J., Smith, R., Schooler, J. W.,
⁵⁸⁵ 2009. Experience sampling during fmri reveals default network and execu⁵⁸⁶ tive system contributions to mind wandering. Proceedings of the National
⁵⁸⁷ Academy of Sciences 106 (21), 8719–8724.

Constantinescu, A. O., OâĂŹReilly, J. X., Behrens, T. E., 2016. Organizing
conceptual knowledge in humans with a gridlike code. Science 352 (6292),
1464–1468.

⁵⁹¹ Diaz, B. A., Van Der Sluis, S., Benjamins, J. S., Stoffers, D., Hardstone, R.,
⁵⁹² Mansvelder, H. D., Van Someren, E. J., Linkenkaer-Hansen, K., 2014. The
⁵⁹³ arsq 2.0 reveals age and personality effects on mind-wandering experiences.
⁵⁹⁴ Frontiers in psychology 5.

⁵⁹⁵ Ding, Z., Xu, R., Bailey, S. K., Wu, T.-L., Morgan, V. L., Cutting, L. E.,
⁵⁹⁶ Anderson, A. W., Gore, J. C., 2016. Visualizing functional pathways in the
⁵⁹⁷ human brain using correlation tensors and magnetic resonance imaging.
⁵⁹⁸ Magnetic resonance imaging 34 (1), 8–17.

Eichenbaum, H., 1993. Memory, amnesia, and the hippocampal system. MIT
 press.

Eichenbaum, H., 2013. Memory on time. Trends in cognitive sciences 17 (2), 81–88.

Eichenbaum, H., Cohen, N. J., 2014. Can we reconcile the declarative memory and spatial navigation views on hippocampal function? Neuron 83 (4),
764–770.

Ellamil, M., Fox, K. C., Dixon, M. L., Pritchard, S., Todd, R. M., Thompson,
 E., Christoff, K., 2016. Dynamics of neural recruitment surrounding the
 spontaneous arising of thoughts in experienced mindfulness practitioners.
 NeuroImage.

Engert, V., Smallwood, J., Singer, T., 2014. Mind your thoughts: Associations between self-generated thoughts and stress-induced and baseline
levels of cortisol and alpha-amylase. Biological psychology 103, 283–291.

Feinberg, D. A., Moeller, S., Smith, S. M., Auerbach, E., Ramanna, S.,
Glasser, M. F., Miller, K. L., Ugurbil, K., Yacoub, E., 2010. Multiplexed
echo planar imaging for sub-second whole brain fmri and fast diffusion
imaging. PloS one 5 (12), e15710.

Gawryluk, J. R., Mazerolle, E. L., D'Arcy, R. C., 2014. Does functional mri
detect activation in white matter? a review of emerging evidence, issues,
and future directions. Frontiers in neuroscience 8, 239.

Golchert, J., Smallwood, J., Jefferies, E., Seli, P., Huntenburg, J. M., Liem,
F., Lauckner, M. E., Oligschläger, S., Bernhardt, B. C., Villringer, A.,
et al., 2016. Individual variation in intentionality in the mind-wandering
state is reflected in the integration of the default-mode, fronto-parietal,
and limbic networks. NeuroImage.

Gorgolewski, K. J., Lurie, D., Urchs, S., Kipping, J. A., Craddock, R. C.,
Milham, M. P., Margulies, D. S., Smallwood, J., 2014. A correspondence
between individual differences in the brain's intrinsic functional architecture and the content and form of self-generated thoughts. PloS one 9 (5),
e97176.

Greicius, M. D., Krasnow, B., Reiss, A. L., Menon, V., 2003. Functional
connectivity in the resting brain: a network analysis of the default mode
hypothesis. Proceedings of the National Academy of Sciences 100 (1), 253–
258.

- Gusnard, D. A., Raichle, M. E., 2001. Searching for a baseline: functional
 imaging and the resting human brain. Nature Reviews Neuroscience 2 (10),
 685–694.
- Hassabis, D., Maguire, E. A., 2007. Deconstructing episodic memory with
 construction. Trends in cognitive sciences 11 (7), 299–306.
- Huijbers, W., Pennartz, C. M., Cabeza, R., Daselaar, S. M., 2011. The
 hippocampus is coupled with the default network during memory retrieval
 but not during memory encoding. PLoS One 6 (4), e17463.
- Jbabdi, S., Behrens, T. E., Smith, S. M., 2010. Crossing fibres in tract-based spatial statistics. Neuroimage 49 (1), 249–256.
- Jones, D. K., 2004. The effect of gradient sampling schemes on measures derived from diffusion tensor mri: a monte carlo study[†]. Magnetic Resonance in Medicine 51 (4), 807–815.
- Jones, D. K., Knösche, T. R., Turner, R., 2013. White matter integrity, fiber count, and other fallacies: the do's and don'ts of diffusion mri. Neuroimage 73, 239–254.
- Kane, M. J., Brown, L. H., McVay, J. C., Silvia, P. J., Myin-Germeys, I.,
 Kwapil, T. R., 2007. For whom the mind wanders, and when an experiencesampling study of working memory and executive control in daily life.
 Psychological science 18 (7), 614–621.
- Killingsworth, M. A., Gilbert, D. T., 2010. A wandering mind is an unhappy
 mind. Science 330 (6006), 932–932.
- Klinger, E., Cox, W. M., 1987. Dimensions of thought flow in everyday life.
 Imagination, Cognition and Personality 7 (2), 105–128.
- Levinson, D. B., Smallwood, J., Davidson, R. J., 2012. The persistence of
 thought evidence for a role of working memory in the maintenance of taskunrelated thinking. Psychological Science 23 (4), 375–380.
- Maguire, E. A., Intraub, H., Mullally, S. L., 2015. Scenes, spaces, and
 memory traces what does the hippocampus do? The Neuroscientist,
 1073858415600389.

Margulies, D. S., Ghosh, S. S., Goulas, A., Falkiewicz, M., Huntenburg, J. M.,
Langs, G., Bezgin, G., Eickhoff, S. B., Castellanos, F. X., Petrides, M.,
et al., 2016. Situating the default-mode network along a principal gradient
of macroscale cortical organization. Proceedings of the National Academy
of Sciences, 201608282.

Mason, M. F., Norton, M. I., Van Horn, J. D., Wegner, D. M., Grafton,
S. T., Macrae, C. N., 2007. Wandering minds: the default network and
stimulus-independent thought. Science 315 (5810), 393–395.

McVay, J. C., Kane, M. J., 2009. Conducting the train of thought: working
memory capacity, goal neglect, and mind wandering in an executive-control
task. Journal of Experimental Psychology: Learning, Memory, and Cognition 35 (1), 196.

McVay, J. C., Kane, M. J., Kwapil, T. R., 2009. Tracking the train of thought
from the laboratory into everyday life: An experience-sampling study of
mind wandering across controlled and ecological contexts. Psychonomic
bulletin & review 16 (5), 857–863.

Medea, B., Karapanagiotidis, T., Konishi, M., Ottaviani, C., Margulies, D.,
Bernasconi, A., Bernasconi, N., Bernhardt, B. C., Jefferies, E., Smallwood,
J., 2016. How do we decide what to do? resting-state connectivity patterns
and components of self-generated thought linked to the development of
more concrete personal goals. Experimental brain research, 1–13.

Mori, S., Wakana, S., Van Zijl, P. C., Nagae-Poetscher, L., 2005. MRI atlas of human white matter. Vol. 16. Am Soc Neuroradiology.

Moscovitch, M., Cabeza, R., Winocur, G., Nadel, L., 2016. Episodic memory and beyond: the hippocampus and neocortex in transformation. Annual review of psychology 67, 105–134.

Murphy, K., Birn, R. M., Handwerker, D. A., Jones, T. B., Bandettini, P. A.,
2009. The impact of global signal regression on resting state correlations:
are anti-correlated networks introduced? Neuroimage 44 (3), 893–905.

Nichols, T. E., Holmes, A. P., 2002. Nonparametric permutation tests for
 functional neuroimaging: a primer with examples. Human brain mapping
 15 (1), 1–25.

- ⁶⁹⁶ O'Keefe, J., Burgess, N., et al., 1996. Geometric determinants of the place ⁶⁹⁷ fields of hippocampal neurons. Nature 381 (6581), 425–428.
- O'keefe, J., Nadel, L., 1978. The hippocampus as a cognitive map. Vol. 3.
 Clarendon Press Oxford.
- Patterson, K., Nestor, P. J., Rogers, T. T., 2007. Where do you know what
 you know? the representation of semantic knowledge in the human brain.
 Nature Reviews Neuroscience 8 (12), 976–987.
- Peer, M., Salomon, R., Goldberg, I., Blanke, O., Arzy, S., 2015. Brain system for mental orientation in space, time, and person. Proceedings of the National Academy of Sciences 112 (35), 11072–11077.
- Poerio, G. L., Totterdell, P., Miles, E., 2013. Mind-wandering and negative mood: Does one thing really lead to another? Consciousness and cognition 22 (4), 1412–1421.
- Raichle, M. E., 2015. The brain's default mode network. Annual review of
 neuroscience 38, 433–447.
- Raichle, M. E., MacLeod, A. M., Snyder, A. Z., Powers, W. J., Gusnard,
 D. A., Shulman, G. L., 2001. A default mode of brain function. Proceedings of the National Academy of Sciences 98 (2), 676–682.
- Ruby, F. J., Smallwood, J., Engen, H., Singer, T., 2013a. How self-generated
 thought shapes mood-the relation between mind-wandering and mood depends on the socio-temporal content of thoughts. PLoS One 8 (10), e77554.
- Ruby, F. J., Smallwood, J., Sackur, J., Singer, T., 2013b. Is self-generated
 thought a means of social problem solving? Frontiers in psychology 4.
- Schacter, D. L., Addis, D. R., Buckner, R. L., 2007. Remembering the past
 to imagine the future: the prospective brain. Nature Reviews Neuroscience
 8 (9), 657–661.
- Schooler, J. W., 2002. Re-representing consciousness: Dissociations between
 experience and meta-consciousness. Trends in cognitive sciences 6 (8), 339–
 344.

Seli, P., Carriere, J. S., Levene, M., Smilek, D., 2013. How few and far
between? examining the effects of probe rate on self-reported mind wandering. Frontiers in psychology 4, 430.

Sestieri, C., Corbetta, M., Romani, G. L., Shulman, G. L., 2011. Episodic
memory retrieval, parietal cortex, and the default mode network: functional and topographic analyses. The Journal of neuroscience 31 (12),
4407-4420.

Smallwood, J., Beach, E., Schooler, J. W., Handy, T. C., 2008. Going awol
in the brain: Mind wandering reduces cortical analysis of external events.
Journal of cognitive neuroscience 20 (3), 458–469.

Smallwood, J., Fitzgerald, A., Miles, L. K., Phillips, L. H., 2009a. Shifting moods, wandering minds: negative moods lead the mind to wander.
Emotion 9 (2), 271.

Smallwood, J., Gorgolewski, K. J., Golchert, J., Ruby, F. J., Engen, H. G.,
Baird, B., Vinski, M. T., Schooler, J. W., Margulies, D. S., 2013a. The
default modes of reading: modulation of posterior cingulate and medial
prefrontal cortex connectivity associated with comprehension and task focus while reading. Frontiers in human neuroscience 7.

Smallwood, J., Karapanagiotidis, T., Ruby, F., Medea, B., de Caso, I., Konishi, M., Wang, H.-T., Hallam, G., Margulies, D. S., Jefferies, E., 2016.
Representing representation: Integration between the temporal lobe and
the posterior cingulate influences the content and form of spontaneous
thought. PloS One 11 (4), e0152272.

- Smallwood, J., Nind, L., O'Connor, R. C., 2009b. When is your head at?
 an exploration of the factors associated with the temporal focus of the
 wandering mind. Consciousness and cognition 18 (1), 118–125.
- Smallwood, J., Ruby, F. J., Singer, T., 2013b. Letting go of the present:
 mind-wandering is associated with reduced delay discounting. Conscious ness and cognition 22 (1), 1–7.

Smallwood, J., Schooler, J. W., 2015. The science of mind wandering: empirically navigating the stream of consciousness. Annual review of psychology
66, 487–518.

Smith, S. M., 2002. Fast robust automated brain extraction. Human brain
mapping 17 (3), 143–155.

- ⁷⁵⁹ Smith, S. M., Jenkinson, M., Johansen-Berg, H., Rueckert, D., Nichols, T. E.,
- ⁷⁶⁰ Mackay, C. E., Watkins, K. E., Ciccarelli, O., Cader, M. Z., Matthews,
- P. M., et al., 2006. Tract-based spatial statistics: voxelwise analysis of multi-subject diffusion data. Neuroimage 31 (4), 1487–1505.
- Smith, S. M., Jenkinson, M., Woolrich, M. W., Beckmann, C. F., Behrens,
 T. E., Johansen-Berg, H., Bannister, P. R., De Luca, M., Drobnjak, I.,
 Flitney, D. E., et al., 2004. Advances in functional and structural mr image
 analysis and implementation as fsl. Neuroimage 23, S208–S219.
- Smith, S. M., Nichols, T. E., 2009. Threshold-free cluster enhancement: addressing problems of smoothing, threshold dependence and localisation in
 cluster inference. Neuroimage 44 (1), 83–98.
- Squire, L. R., Stark, C. E., Clark, R. E., 2004. The medial temporal lobe*.
 Annu. Rev. Neurosci. 27, 279–306.
- Stawarczyk, D., D'Argembeau, A., 2015. Neural correlates of personal goal
 processing during episodic future thinking and mind-wandering: An ale
 meta-analysis. Human brain mapping 36 (8), 2928–2947.
- Stawarczyk, D., Majerus, S., Maquet, P., D'Argembeau, A., 2011. Neural
 correlates of ongoing conscious experience: both task-unrelatedness and
 stimulus-independence are related to default network activity. PloS one
 6 (2), e16997.
- Strange, B. A., Witter, M. P., Lein, E. S., Moser, E. I., 2014. Functional
 organization of the hippocampal longitudinal axis. Nature Reviews Neuroscience 15 (10), 655–669.
- Tusche, A., Smallwood, J., Bernhardt, B. C., Singer, T., 2014. Classifying
 the wandering mind: revealing the affective content of thoughts during
 task-free rest periods. Neuroimage 97, 107–116.
- Tzourio-Mazoyer, N., Landeau, B., Papathanassiou, D., Crivello, F., Etard,
 O., Delcroix, N., Mazoyer, B., Joliot, M., 2002. Automated anatomical
 labeling of activations in spm using a macroscopic anatomical parcellation
 of the mni mri single-subject brain. Neuroimage 15 (1), 273–289.

- Woolrich, M., 2008. Robust group analysis using outlier inference. Neuroimage 41 (2), 286–301.
- Woolrich, M. W., Behrens, T. E., Beckmann, C. F., Jenkinson, M., Smith,
 S. M., 2004. Multilevel linear modelling for fmri group analysis using
 bayesian inference. Neuroimage 21 (4), 1732–1747.
- Yeo, B. T., Krienen, F. M., Sepulcre, J., Sabuncu, M. R., Lashkari, D.,
 Hollinshead, M., Roffman, J. L., Smoller, J. W., Zöllei, L., Polimeni, J. R.,
 et al., 2011. The organization of the human cerebral cortex estimated by
 intrinsic functional connectivity. Journal of neurophysiology 106 (3), 1125–
 1165.

799 Figure captions

Figure 1. Overview of analysis pipeline.

The upper panel describes the analysis steps which allow the dimensions that underlie the trial level experience sampling data to be calculated. The middle panel describes how the functional connectivity maps for our regions of interest are calculated from the resting state functional Magnetic Resonance Imaging data (MRI). The lower panel describes how the fractional anisotropy maps are calculated from the whole-brain diffusion MRI data.

807

Figure 2. Identifying the relationship between structural connectivity and the contents of spontaneous thought.

i: The upper panel shows the results of the whole-brain diffusion MRI anal-810 vsis. The clusters where a significant increase in fractional anisotropy was 811 found for participants engaging more in mental time travel are indicated in 812 red and are overlaid on the mean fractional anisotropy skeleton. Results 813 were thresholded at a Family-Wise Error (FWE) corrected p < 0.05 using 814 Threshold-Free Cluster Enhancement. The lower panel presents these signif-815 icant clusters along with the probabilistic streamline that was found to pass 816 through them. The streamline was thresholded using Threshold-Free Cluster 817 Enhancement at p < 0.05, FWE-corrected. 818

ii: This panel demonstrates the overlap of the probabilistic streamline with the right fornix, the right corticospinal tract, and the right inferior frontooccipital fasciculus.

Acronyms: FA - fractional anisotropy , MTT+ - increased mental time travel, PS - probabilistic streamline, rFX - right Fornix, rCST - right corticospinal tract, riFOF - right inferior fronto-occipital fasciculus.

825

Figure 3. Identifying the grey matter regions connected to the temporolimbic white matter substrate of mental time travel.

The box plots in the upper panel show the connection probability of each one 828 of the 116 grey matter volumes of the Automated Anatomical Labelling atlas 829 with the white matter substrate of mental time travel found in our whole-830 brain tractography analysis. In the lower panel, the volumes are presented 831 with each region coloured according to its average connection probability 832 among participants. It is clear in both panels that the right hippocampus 833 has the highest number of streamlines connecting it to the cluster obtained 834 in the prior step of our analysis. 835

Figure 4. Determining the link between the functional connectivity of the hippocampus and the content of spontaneous thought.

836

i: The group-level functional connectivity of the right hippocampus (left), the region showing a stronger temporal connectivity with it for individuals with higher MTT (middle) and the scatter plot (right) showing the average beta values extracted from this region plotted against the mental time travel scores for each participant. Maps were thresholded at the whole-brain level with a cluster-forming threshold of z > 3.1 and a FWE corrected cluster significance level of p < 0.05.

ii. The upper panel shows the regions with increased functional connectivity 846 with the right hippocampus for individuals with higher MTT, when each 847 sample is analysed separately (top left and top middle), as well as their 848 overlap (top right). The lower panel presents an axial slice of the overlap 840 cluster and the two scatter plots from each sample that illustrate the average 850 beta values extracted from this region plotted against the mental time travel 851 scores for each participant. It is apparent that this relationship is consistent 852 across both datasets. Maps were thresholded at the whole-brain level with a 853 cluster-forming threshold of z > 2.3 and a FWE corrected cluster significance 854 level of p < 0.05. 855

iii: The relationship between the default mode network as defined in the Yeo et al. (2011) study and the group-level functional connectivity of the mental time travel conjunction cluster produced through the analysis of an independent dataset. Maps were thresholded at the whole-brain level with a cluster-forming threshold of z > 3.1 and a FWE corrected cluster significance level of p < 0.05.

Acronyms: MTT - mental time travel, DMN - default mode network.

Table 1. Experience sampling questions used in this experiment.			
Dimension	Question (My thoughts)	Left	Right
Task	were focused on the task I was performing	Not at all	Completely
Future	involved future events	Not at all	Completely
Past	involved past events	Not at all	Completely
Self	involved myself	Not at all	Completely
Other	involved other people	Not at all	Completely
Emotion	The content of was	Negative	Positive
Images	\dots were in the form of images	Not at all	Completely
Words	were in the form of words	Not at all	Completely
Intrusive	were intrusive	Not at all	Completely
Detail	were vague and non-specific	Not at all	Completely

Table 1: Experience sampling questions used in this experiment.

Table 2: Clusters showing a significant association between the MTT weights and the FA of the whole-brain, corrected for multiple comparisons with a family-wise error rate of p <0.05.

Cluster size (voxels)	Cluster centre of gravity X,Y,Z (mm)
481	25, -18, 9
172	32, -22, -5
8	27, -31, -3