# PNAS www.pnas.org

| 1<br>2<br>3  |  |  |  |
|--|--|--|--|
| 4  |  |  |  |
| 5<br>6<br>7<br>8   | Supplementary Information for  |  |  |
| 9  | Transcranial Magnetic Stimulation Induced Perturbations of Resting-  |  |  |
| 10   | State-Networks are Reproducible Markers of Causal Network  |  |  |
| 11   | Dynamics   |  |  |
| 12   |  |  |  |
| 13   | Recep A. Ozdemir <sup>1</sup> , Ehsan Tadayon <sup>1</sup> , Pierre Boucher <sup>1</sup> , Davide Momi <sup>1</sup> , Kelly A.         |  |  |
| 14   | Karakhanyan <sup>1</sup> , Michael D. Fox <sup>1,2</sup> , Mark Halko <sup>1</sup> , Alvaro Pascual-Leone <sup>*1,3</sup> , Mouhsin M. |  |  |
| 15   | Shafi <sup>*1</sup> , Emiliano Santarnecchi <sup>*1,4</sup>  |  |  |
| 16<br>17<br>18   | Emiliano Santarnecchi<br>Email: <u>esantarn@bidmc.harvard.edu</u>  |  |  |
| 19<br>20<br>21<br>22<br>23<br>24<br>25<br>26<br>27                   | This PDF file includes:<br>Supplementary methods and Results<br>Figures S1 to S4<br>Legends for Movies S1<br>SI References             |  |  |
| 28<br>29   | Other supplementary materials for this manuscript include the following:   |  |  |
| 30<br>31<br>32<br>33<br>34<br>35<br>36<br>37<br>38<br>39<br>40<br>41 | Movie SV_1<br>Movie SV_2   |  |  |

#### **1** Supplementary Methods

2

### 3 Resting-state fMRI data and preprocessing

4 Briefly, structural T1-weighted processing included surface reconstruction using 5 Freesurfer, spatial normalization, brain extraction and tissue segmentation (white-matter, 6 gray-matter and cerebrospinal fluid (CSF)). Functional data were slice time corrected and 7 motion-corrected and registered to the T1-weighted structural MRI using boundary-based 8 registration with six degrees of freedom. CompCor approach (with six temporal and six 9 anatomical components) was deployed for physiological noise regressions, and the time 10 series were bandpass filtered (0-008 - 0.08). The time-series were resampled onto 11 fsaverage4 for creating individualized networks.

12

# 13 Generating individual networks

First, group-average functional networks are morphed into the subject's cortical surface using surface-based registration. Then, using the subject's preprocessed restingtate fMRI data, the vertices are re-assigned to one of the 18 networks in an iterative procedure based on the correlation between each vertex and the average time-course for each network while considering the variability of each network across subjects (i.e. confidence map). Networks #15-17 were combined to get DMN, and Networks #5 and #6 were combined to get DAN (SI Appendix, Fig. S4).

21

# 22 Sandia Matrices

23 SM is a recently developed PC-based, matrix-like abstract reasoning task resembling and expanding the Raven's Matrices (Raven and Court, 1998). Compared to the original 24 25 Raven stimuli, the SM tool includes multiple sets of validated stimuli which allow for 26 repeated measurement of gf, while allowing for sensitive recording of response times 27 (Santarnecchi et al., 2016, 2013). Each matrix is composed of a 3x3 grid, with each cell in 28 the grid containing a set of shapes. A blank cell is present in position 3-3 of the grid 29 (bottom right). Participants were required to complete the matrix using one of eight options. 30 Participants responded by pressing the corresponding key on a keyboard (1-8). A maximum 31 of 60 seconds was given for each matrix to be completed, after which the next matrix 32 appeared. Two different sets of matrices were presented: Relational matrices (gf relational) 33 and Logic matrices (gf logical). Relational matrices could be solved by capturing variations 34 in the features of shapes across cells in the grid, (i.e., color, size, orientation, number, 35 shape), with some features staying constant while others vary. Logic matrices required 36 participants to perform logical operations across the matrix (i.e., conjunction, disjunction, 37 or exclusive disjunction, known as AND, OR and XOR functions, respectively), similarly 38 to the conditional inferences discussed above. Logic and relational matrices were presented 39 in a randomized order, and the interstimulus interval was 5 seconds. Additionally, before 40 starting the first block, participants performed a short training exercise to familiarize with 41 the stimuli and reduce possible novelty effects in the first block. Before starting the training 42 block, brief verbal instructions were given to participants explaining the task. Following 1 training participants completed 12 trials of both Relational and Logic matrices presented

2 in a randomized order (24 matrices in total, with each matrix appearing for 60 sec),

3

# 4 TOPF

5 The TOPF provided an estimate of premorbid intellectual abilities. Participants were 6 requested to read aloud a total of 70 words, focusing on pronunciation. An estimated 7 premorbid IQ score was calculated based on the number and type of errors. The test has 8 been used in multiple cognitive studies (Alves et al., 2013) and clinical trials (Berg et al., 9 2016), providing a brief, validated estimate of IQ.

10

11 SRTT

12 A visual cue was presented at four different positions, configured horizontally at the 13 center of a computer screen. Subjects were asked to press the response button 14 corresponding to the position of the visual cue as quickly as possible. The visual cues were 15 presented either in a repeating sequence or in a random order. In the sequenced trials, the 16 location of the stimulus followed a fixed order, while in random trials the stimuli were 17 presented in random locations. When the task was finished, a questionnaire was fulfilled where participants were asked to indicate whether the visualization pattern was based on a 18 19 repeated sequence or not. Subjects were unware about the questionnaire before the task.

20

# 21 EMG Data Processing

EMG data were amplified and digitized using a Powerlab 4/25T data acquisition system (ADInstruments) at a sampling rate of 4000 Hz (bandpass filtered at 10Hz to 2000Hz). EMG signals were continuously streamed by using LabChart software (LabChart 8.0) to monitor MEPs and epochs were recorded with a 150ms window length covering from 50 ms before to 100 ms after TMS pulse.

27

# 28 *E-Field modeling*

29 The e-fields were simulated using SIMNIBS (v2.1) software. The head models 30 were reconstructed using *mri2mesh* pipeline, which creates tetrahedral meshes consisting 31 of white-matter (WM), gray-matter (GM), CSF, skull and scalp. We used the default 32 conductivity values of 0.126 S/m for WM, 0.275 S/m for GM, 1.654 S/m for CSF, 0.01 S/m 33 for the skull, and 0.465 S/m for scalp. We measured dI/dt from the Magventure Machine 34 based on the stimulus intensity. The calculated E-fields were normalized by the maximum 35 value of E-fields for each stimulation target. The normalized E-field maps were mapped to 36 the fsaverage common template and averaged across subjects for visualization purposes.

37

# 38 EEG data Processing

First, the two separate blocks of 60 trials were merged to create a single block of trials. The merged trials were then segmented into 3000ms epochs, each starting before (pre-stimulus) and ending 2000ms (post-stimulus) following TMS pulse,

1 respectively. Baseline correction was performed by subtracting the mean pre-stimulus (-2 900 to -100) signal amplitude from the rest of the epoch in each channel. Following 3 baseline correction, data were visually inspected to identify extremely noisy channels with 4 high amplitudes  $(3 \pm 2 \text{ channels were deleted on average; range 0-6 out of 63. See Figure$ 5 S1 step 3 for a sample). Zero-padding between -2ms and 14ms time range were then applied 6 to remove the early TMS pulse artifact from the EEG data. All zero padded epochs were 7 then tagged based on voltage ( $\geq 100 \,\mu$ V), kurtosis ( $\geq 3$ ), and joint probability (Single 8 channel-based threshold  $\geq$  3.5sd; All channel-based threshold  $\geq$  5sd) metrics to identify 9 excessively noisy epochs. Visual inspection was performed on the tagged epochs for the 10 final decision for the removal of noisy epochs ( $24 \pm 10$  epochs were deleted on average; range 2-44 out of 120; See Figure S1 steps 4 and 5 for a sample). Next, an initial round of 11 12 fast independent component analysis (fICA) (43) was performed to exclusively identify 13 and remove components with early TMS evoked high amplitude electrode and EMG 14 artifacts ( $1\pm 1$  components were removed; range 0-3 out of 60, See Figure S1 step 6 for a 15 sample). Prior to the first round of fICA, the EEG data were reduced into 60 dimensions 16 by using principal component analyses (PCA) to minimize overfitting and noise 17 components. After the first round of fICA, the EEG data were interpolated for previously zero-padded time window around TMS pulse (-2ms to 14ms) using linear interpolation, 18 band pass filtered using a forward-backward 4<sup>th</sup> order butterwoth filter from 1 to 100Hz, 19 20 notch filtered between 57 and 63Hz, and referenced to global average (See Figure S1 step 21 7 for a sample data set). Subsequently, the data were reduced into 57 dimensions using 22 PCA, and a second round of fICA was run to manually remove all remaining artifact 23 components<sup>4</sup> including eve movement/blink, muscle noise (EMG), single electrode noise, 24 TMS evoked muscle, cardiac beats (EKG), as well as auditory evoked potentials ( $22\pm 6$ 25 components were removed; range 18-28 out of 37, See Figure S1 step 8 for topography, 26 time series and power spectrum of representative artifact components). In both rounds of 27 fICA, a semi-automated artifact detection algorithm incorporated into the open source 28 TMS-EEG Signal Analyzer (TESA v0.1.0-beta) extension for EEGLAB was used to 29 classify and visually inspect components based on their frequency, activity power 30 amplitude, time spectrum, scalp topography, and course (44)31 (http://nigelrogasch.github.io/TESA/). Finally, the data were low pass filtered with a 4<sup>th</sup> 32 order Butterworth filter at 50Hz, and interpolated for missing/removed channels using 33 spherical interpolation (See Figure S1 step 9 for a sample cleaned data set with topography 34 of selected TMS evoked potentials).

35

## 36 *GMFP and LMFP*

Global cortical activation level following TMS of DMN and DAN was measured
 with Global Mean Field Power (GMFP) analyses on TEPs using the following equation.

39

40 
$$GMFP(t) = \sqrt{\left\{\frac{\sum_{i}^{k} (\text{Vi}(t) - V_{mean(t)})2}{K}\right\}}$$

1 Where  $V_i(t)$  is the voltage at electrode *i* at a certain point in time,  $V_{\text{mean}(t)}$  is the mean of 2 instantaneous TEP across electrodes, and K is the number of electrodes. The GMFP can be 3 used to measure the global brain response to TMS (Lehmann and Skrandies, 1980; Komssi 4 et al, 2004). A Local Mean Field Power (LMFP) analysis was also performed to identify 5 local changes in cortical excitability induced by TMS of DMN and DAN. Region of 6 Interests (ROIs) were determined separately for each hemisphere from the selected 7 channels over frontal (Right Frontal 'RF' electrodes: FC2, FC4, FC6, F2, F4, F6 and Left 8 Frontal 'RF' electrodes: FC1, FC3, FC5, F1, F3, F5), motor (Right Motor 'RM' electrodes: 9 C2, C4, C6 and Left Motor 'LM' electrodes: C1, C3, C5), parietal (Right Parietal 'RP' 10 electrodes: CP2, CP4, CP6, P2, P4, P6 and Left Parietal 'LP' electrodes: CP1, CP3, CP5, P1, P3, P5) and visual cortices (Right Visual 'RV' electrodes: PO4, PO8, O2 and Left 11 12 Visual 'LV' electrodes: PO3, PO7, O1). LMFP was calculated as root-mean-square value 13 of the ROI electrodes for each time point (Casarotto et al., 2013).

14 TEP time series were used to test reproducibility of TMS evoked global and local 15 cortical responses in electrode space.

16

#### 17 Supplementary Results

### 18 DMN stimulation induces higher global cortical activation in the resting brain.

19 GMFP time series for a representative subject, comparison of group mean GMFP 20 time series with standard deviations for DMN and DAN stimulations, and individual sum 21 of GMFP values, computed as the total amount of cortical activation from 0 to 400ms post 22 TMS, with trend lines were provided in supplementary figure S2A<sub>1</sub>, S2A<sub>2</sub> and S2A<sub>3</sub> 23 respectively. We first computed cluster-based permutation (permutation n=10000) paired 24 t-test statistics (Maris and Oostenweld, 2007) on GMFP time series at each time point from 25 -200 to 400ms to compare temporal evolution in global cortical activation between DMN 26 and DAN stimulation. Although DMN stimulation has higher values (Figure S2A<sub>2</sub>), 27 cluster-based permutation pairwise comparisons did not reveal significant differences (p >28 0.05) at any time point between DMN and DAN. However, pairwise comparison for 29 individual sum of GMFP values showed that DMN simulation resulted in increased global 30 cortical excitability (p = 0.03) within 400ms window following TMS, when compared to 31 DAN stimulation (Figure S2A<sub>3</sub>), suggesting that, at the group level, total amount of global 32 cortical activation was higher following DMN stimulation in the resting brain. Given high 33 inter-individual variability in sum of GMFP values (See shaded regions in figure S2A<sub>2</sub>) and a recent study (Jannati et al., 2017) showing the BDNF polymorphism as potential 34 35 confounder in TMS induced responses, we split our participants into two groups based on 36 their BDNF polymorphism (Val66Met vs Val66Val), and compared their GMFP values 37 both for DAN and DMN stimulation (Figure S2). Independent sample t-tests revealed no 38 significant differences between groups (p > 0.05). Additionally, we ran control analyses to 39 further examine whether higher GMFP values following DMN stimulation is confounded 40 with possible different muscle activation patterns across stimulation conditions due to more 41 medial or lateral targeted brain regions. Given the proximity of DMN stimulation targets 42 to scalp/neck muscles we carefully inspected the systematic presence of stronger muscle 43 activations following DMN stimulation which might potentially startle the subjects and 44 thus potentially induce saliency-related Event-Related-Potentials. An additional cycle of

1 ICA analyses was ran using all session-1 data sets for each participant, by systematically 2 keeping the TMS induced muscle components while removing all other components from 3 the data. Supplementary figure S3 below shows examples of TMS induced muscle 4 components for a representative subject (and for both stimulation conditions) that were kept in the data. We then quantified the amplitude of TMS induced muscle responses with 5 GMFP analyses (figure S3B). Analysis of time series (Figure S3B, left panel) and overall 6 7 response (figure S3B, right panel) showed highly similar and non-significant differences 8 between stimulation conditions (p > 0.05), suggesting that there is no systematic presence of significantly stronger muscle responses during DMN stimulation. 9

10 Similar to GMFP time series analyses, permutation based pairwise comparisons for 11 LMFP time series also failed to reveal significant differences at any time point for all ROIs 12 (p > 0.05), although the DMN time series were higher in general (Supplementary Figure 13 S4B<sub>2</sub>). Pairwise comparisons for individual sum of LMFP values did not show any 14 significant differences (p > 0.05) between DMN and DAN stimulation for the total amount 15 of local cortical activations at any ROIs (Figure S4B<sub>3</sub>).

16

# 17 Test re-test reliability of TEPs

18 To test the reproducibility of TMS evoked cortical activation dynamics, we first 19 generated grand mean TEP time series, computed as the mean time series of all 20 participants, for each EEG channel across stimulation conditions (DMN vs DAN) and 21 sessions (Session-1 vs Session-2) (Figure S6A shows TEPs for DAN and Figure S6B 22 shows TEPs for DMN). We then computed the Pearson correlations across grand mean 23 TEP time series at every 100ms windows from -100ms to 700ms after TMS. Topographical 24 evolution of these correlations are provided in figure S6C for each 100ms window. Scalp map topographies for pre-stimulus window (-100 to 0ms) showed predominantly low 25 26 negative correlations (r < 0.5), in both stimulation condition across sessions. For the DAN 27 stimulation while increased positive correlations were mainly observed over bilateral 28 prefrontal and left medial parietal cortices (r range: 0.11-0.89) 100ms following TMS, 29 highest positive correlations across DAN sessions were achieved between 101-200ms and 30 201-300ms windows for the entire scalp (r range: 0.56-0.99), except electrodes over central 31 parietal region close to stimulation site (CP1: r=0.29, and CPz: r=0.44). Positive 32 correlations notably diminished after 300ms and returned to pre-stimulus values 500-33 600ms following TMS. For the DMN stimulation positive correlations mainly rose 34 bilaterally over temporal-parietal, visual, motor and pre-motor cortices in the first 100ms 35 window (r range: 0.64-0.93) and peaked between 101-200ms and 201-300ms windows 36 following TMS (r range: 0.79-0.99), except in temporal parietal electrodes that are closer 37 to stimulation site (C6: r=0.37, and CP6: r=0.51). Contrary to the DAN stimulation, 38 positive correlations following DMN stimulation survived up to 600ms window especially 39 over frontal cortices and returned to pre-stimulus values 600-700ms following TMS. 40 Comparison of global cortical correlation values, computed as the grand mean of all scalp 41 electrodes, between the DAN and DMN stimulation revealed significantly higher 42 correlations for the DMN scalp maps at 301-400ms (p = 0.02), 401-500ms (p = 0.03) and 43 501-600ms (p = 0.01) windows, suggesting prolonged reproducibility in TEPs following 44 DMN stimulation (Figure S6D).

#### **Supplementary Figures**

# 



8 9

Supplementary Figure 1. EEG Signal Preprocessing: This supplementary figure shows major steps of EEG data cleaning procedures on a sample data set (panel A and B), and resulting signal transformation showing evolution of TMS evoked potentials (TEPs), with final scalp topography of a selected TMS evoked potential peak (Step 9). See methods for details.





Supplementary Figure 2. TMS targets for DAN (green circle) and DMN (red circle) were shown in Panel A for a representative subject. TEPs (black lines representing each EEG electrode) and grand mean (solid colored lines) of all electrodes with resulting GMFP time series (colored areas under the integrated grand mean curves) were shown in Panel A1. Group means (solid lines) and standard deviations (shaded regions) of GMFP time series for each network stimulation were shown in panel A<sub>2</sub>. The area-under-the-curve of individual TMS evoked GMFPs (colored dots) and grand means (solid colored lines) of DAN and DMN stimulation along with between network comparison statistics were shown in Panel A3 (Individual trends in TMS evoked GMFPs, as a function of network stimulation, were shown with solid black lines connecting each individual's GMFP values across stimulation conditions. Note that DMN stimulation induces significantly (p=0.04) higher global electro-cortical activity at the group level, when compared to DAN stimulation).



Time (ms)
Supplementary Figure 3: TMS induced muscle components. (A) Examples of TMS induced muscle components from a representative subject for both stimulation conditions (DAN: left panel, DMN: right panel). (B) Group averaged GMFP time series (left panel) for TMS induced muscle components (DAN: Blue solid line, DMN: red solid line with corresponding shaded regions representing standard error of mean), and sum of GMFP activity (right panel) for 100ms following TMS.



**Supplementary Figure 4.** (A) Color codes of selected EEG electrode groups for LMFP computations in each ROI. Dark colored channels represent frontal (blue), motor (green), parietal (red) and visual (brown) ROIs for LMFPs on the right hemisphere (stimulated), respectively. Corresponding light colored channels represents left hemisphere ROIs for LMFPs. LMFP profiles of stimulated hemisphere for a representative subject, Grand mean LMFP time series (colored solid lines) with standard deviations (color shaded regions), and individual TMS evoked LMFP values of each network stimulation along with between network comparison statistics were shown in Panels A<sub>1</sub>, A<sub>2</sub>, and A<sub>3</sub>, respectively.



Supplementary Figure 5. (A) Individualized Networks. Group-average functional networks are morphed into subject's anatomy and through an iterative process the vertices are assigned to the network that they have the highest correlation with. The individualized networks as well as the individualized DMN and DAN are shown for two representative subjects. (B) Group-based (upper panels) and individual network maps (lower panels) on a common template. (C) Comparison of group averaged current density time series across stimulation conditions extracted both from group-based (upper panel) and individual (lower panels) network maps. (D) Sum of overall network activity 400ms following TMS across stimulation conditions.



**Supplementary Figure 6**: Topographical similarity of TEPs between two identical TMS sessions (Reliability of TEPs): Grand mean TEP time series of EEG electrodes on the right (stimulated) hemisphere following DAN (panel A) and DMN (Panel B) stimulation across sessions (black solid lines show TEPs recorded in visit-1 while gray solid lines shows TEPs recorded in visit 2). Colored time series are expanded and presented on the right side of Panels A and B to show time course of representative channels from frontal (F4), parietal (CP2) and visual (O2) cortices across visits. Topography of test re-test correlations (r values of each EEG electrode) were shown in panel C both for DAN (upper scalp maps) and DMN (lower scalp maps) stimulation with 100ms moving windows before (-100 - 0ms) and after (1 to 700ms) TMS. Panel D show global mean of test re-test correlations at each time window (Color coded to match scalp maps shown in panel C), for DAN (green bars with standard errors) and DMN (red bars with standard errors).



# **References**

| 2  |    |  |
|----|----|--|
| 3  | 1) | Alves, L., Simões, M.R., Martins, C., Freitas, S., Santana, I., 2013. Premorbid IQ   |
| 4  |    | influence on screening tests' scores in healthy patients and patients with cognitive |
| 5  |    | impairment. J. Geriatr. Psychiatry Neurol. 26, 117–126.                              |
| 6  | 2) | Berg, JL., Durant, J., Banks, S.J., Miller, J.B., 2016. Estimates of premorbid       |
| 7  |    | ability in a neurodegenerative disease clinic population: comparing the Test of      |
| 8  |    | Premorbid Functioning and the Wide Range Achievement Test. Clin.                     |
| 9  |    | Neuropsychol. 30, 547–557.   |
| 10 | 3) | Casarotto, S., Canali, P., Rosanova, M., Pigorini, A., Fecchio, M., Mariotti, M.,    |
| 11 |    | Lucca, A., Colombo, C., Benedetti, F., Massimini, M., 2013. Assessing the effects    |
| 12 |    | of electroconvulsive therapy on cortical excitability by means of transcranial       |
| 13 |    | magnetic stimulation and electroencephalography. Brain Topogr. 26, 326–337.          |
| 14 | 4) | Jannati, A., Block, G., Oberman, L.M., Rotenberg, A., Pascual-Leone, A., 2017.       |
| 15 |    | Interindividual variability in response to continuous theta-burst stimulation in     |
| 16 |    | healthy adults. Clin. Neurophysiol. 128, 2268–2278.                                  |
| 17 |    | https://doi.org/10.1016/j.clinph.2017.08.023   |
| 18 | 5) | Raven, J.C., Court, J.H., 1998. Raven's progressive matrices and vocabulary          |
| 19 |    | scales. Oxford pyschologists Press.  |
| 20 | 6) | Santarnecchi, Muller, T., Rossi, S., Sarkar, A., Polizzotto, N.R., Rossi, A., Cohen  |
| 21 |    | Kadosh, R., 2016. Individual differences and specificity of prefrontal gamma         |
| 22 |    | frequency-tACS on fluid intelligence capabilities. Cortex 75, 33–43.                 |
| 23 |    | https://doi.org/10.1016/j.cortex.2015.11.003   |
| 24 | 7) | Santarnecchi, Polizzotto, N.R., Godone, M., Giovannelli, F., Feurra, M., Matzen,     |
| 25 |    | L., Rossi, A., Rossi, S., 2013. Frequency-Dependent Enhancement of Fluid             |
| 26 |    | Intelligence Induced by Transcranial Oscillatory Potentials. Curr. Biol. 23, 1449–   |
| 27 |    | 1453. https://doi.org/10.1016/j.cub.2013.06.022                                      |
| 28 |    |  |