"Evaluating Acquisition Time of rfMRI in the Human Connectome Project for Early Psychosis. How much is enough?"

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Abstract. Resting-state functional MRI (rfMRI) correlates activity across brain regions to identify functional connectivity networks. The Human Connectome Project (HCP) for Early Psychosis has adopted the protocol of the HCP Lifespan Project, which collects 20 min of rfMRI data. However, because it is difficult for psychotic patients to remain in the scanner for long durations, we investigate here the reliability of collecting less than 20 minutes of rfMRI data. Varying durations of data were taken from the full datasets of 11 subjects. Correlation matrices derived from varying amounts of data were compared using the Bhattacharyya distance, and the reliability of functional network ranks was assessed using the Friedman test. We found that correlation matrix reliability improves steeply with longer windows of data up to 11-12 minutes, and ≥ 14 minutes of data produces correlation matrices within the variability of those produced by 18 minutes of data. The reliability of network connectivity rank increases with increasing durations of data, and qualitatively similar connectivity ranks for ≥ 10 minutes of data indicates that 10 minutes of data can still capture robust information about network connectivities.

Keywords: resting state, acquisition time, Bhattacharyya distance.

1 Introduction

Resting-state functional magnetic resonance imaging (rfMRI) can be used to correlate spontaneous fluctuations in the blood oxygen level-dependent (BOLD) signal across regions of the resting brain in order to identify functional connectivity networks [1]. rfMRI is well suited to studying patient populations compared to task-based function-

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al MRI [2] because it does not require participants to perform any tasks, reducing patient burden and expanding the patient pool to those whose cognitive or physical impairment would otherwise preclude them, and has good signal to noise ratio. rfMRI has potential applications in studying group differences, evaluating treatment, and as a diagnostic biomarker [2]. In recent years, it has increasingly been used to investigate changes in functional networks arising from neurological and psychiatric disorders, including stroke, Alzheimer's disease, depression, and schizophrenia [3].

An open question in rfMRI methodology is how much data to acquire. Typical rfMRI studies only acquire 5-10 minutes of data, but the reliability of estimating functional networks improves with increased scan time. However, there are varying recommendations for how many minutes of data are sufficient. While Van Dijk et al. [4] reported that connectivity estimates stabilized at ~5 minutes of rfMRI data, Anderson et al. [5] suggested at least 25 minutes be collected, and recently, Laumann et al. [6] saw further gains in reliability with up to 100 minutes of data. The Washington University-University of Minnesota Human Connectome Project (HCP) [7] collected 60 minutes of rfMRI in four runs of 15 minutes. The relatively long scan time was intended to counteract SNR loss due to the high spatial resolution of the data [8].

Our study collects Connectome-like high quality imaging data in an early psychosis population. We have adopted the protocol of the shorter HCP Pilot Lifespan Project [9], which collects ~60 minutes total of imaging data and includes ~20 minutes of rfMRI data, to make the scan time more tolerable for patients. However, because it is difficult for psychotic patients to remain in the scanner for long durations, we here investigate the reliability of connectivity estimates from less than 20 minutes of rfMRI data.

2 Methods

2.1 Participants

Eleven healthy male participants between the ages of 16 and 35 were recruited from Indiana University, Beth Israel Deaconess Medical Center, Massachusetts General Hospital, and McLean Hospital. The Structured Clinical Interview for DSM-5-RV (SCID) [10] was administered to confirm that subjects did not meet criteria for psychiatric diagnosis. All subjects gave written informed consent.

2.2 Data Acquisition

Imaging data were collected on two Siemens Prisma 3.0 T MRI scanners at Indiana University and Brigham and Women's Hospital with a 32-channel head coil using a modified protocol based on the HCP LifeSpan Pilot project at the University of Minnesota [9]. Structural, functional, and diffusion data were acquired, with a total scan time of about 70 minutes. Structural data included a T1-weighted image (MPRAGE, 0.8 mm³ isotropic resolution, TR/TE=2400/2.22 ms, TI=1000 ms, flip-angle=8°) and a T2-weighted image (SPACE, 0.8 mm³ isotropic resolution, TR/TE=3200/563 ms). Multi-shell diffusion data were acquired four times, twice with anterior-posterior (A-

P) phase encoding and twice with posterior-anterior (P-A) phase encoding (1.5 mm³ isotropic resolution, 92 directions, TR/TE=3230/89.2 ms, multiband acceleration factor=4, b=1500 and 3000). rfMRI data were also acquired four times, twice with A-P phase encoding and twice with P-A phase encoding (EPI, 2.0 mm³ isotropic resolution, TR/TE=800/37 ms, multiband acceleration factor=8). Each functional run consisted of 420 time points, totaling 5.6 minutes. Subjects were instructed to keep their eyes open during all functional scans. In addition, spin echo field maps were acquired to correct for intensity and geometric distortions.

2.3 Data Preprocessing

Structural and functional data were preprocessed with the HCP Minimal Preprocessing Pipelines [11]. The structural pipeline entails registration of the T1w and T2w images, bias field correction, registration to a MNI space template to allow comparison across subjects, reconstruction of white and pial surfaces, and surface registration to a surface atlas. Finally, the data are converted to the Connectivity Informatics Technology Initiative (CIFTI) file format, which combines surface cortical data and volumetric subcortical data in a single "grayordinate" coordinate system.

rfMRI preprocessing includes echo planar imaging (EPI) distortion correction, rfMRI to T1w image registration, motion correction, intensity normalization, and conversion to CIFTI format. The rfMRI data were additionally processed with the Oxford University Centre for Functional MRI of the Brain (FMRIB) group's independent component analysis-based Xnoiseifer - FIX (ICA FIX) [12] pipeline to decompose the BOLD signal and regress out components of physiological noise and noise caused by motion. The resulting time series were then de-meaned and normalized by the standard deviation [13] and concatenated to produce a single time series of 1,680 time points (22.4 minutes).

Contiguous windows of varying durations, ranging from 5.6 to 18 minutes, were randomly extracted from the full 22.4 minute dataset. Durations above 18 minutes were not extracted because the number of different windows that could be obtained would be limited. One hundred windows were extracted for each duration (i.e. 100 of 5.6 minutes, 100 of 6 minutes, etc.). Three *a priori* parcellation schemes (Fig. 1) were then applied to each window of data: the HCP Nature parcellation (360 regions) [14], Yeo's 17-network parcellation (100 regions), and Yeo's 7-network parcellation (8 regions) [15]. Using Pearson's correlation coefficient, region-to-region functional correlation matrices were generated for all parcellations of each window of data. Using the 17-network parcellation only, the mean within-network connectivities were calculated by averaging the correlations found between regions belonging to the same network. Each network was ranked from lowest to highest mean within-network connectivity.

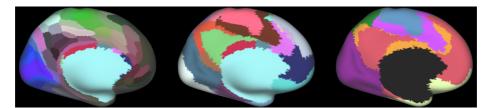


Fig. 1. Different rfMRI parcellation schemes of varying granularity shown on an inflated surface. Left: HCP Nature (360 regions), Center: Yeo 17-network (100 regions), and Right: Yeo 7-network (8 regions).

2.4 Statistical Analysis

Correlation Matrix Reliability. Each duration of data was directly compared to each of the 18-minute windows within each parcellation to determine matrix reliability. Correlation matrices were directly compared using the Bhattacharyya distance D_B [16], which measures the similarity of two probability distributions. The distance D_B for two multivariate normal distributions with means (μ_1 , μ_2) and covariance matrices (Σ_1 , Σ_2) has the form

$$D_B = \frac{1}{8} (\mu_1 - \mu_2)^T \Sigma^{-1} (\mu_1 - \mu_2) + \frac{1}{2} \ln(\frac{|\Sigma|}{\sqrt{|\Sigma_1||\Sigma_2|}}),$$
(1)

where $\Sigma = (\Sigma_1 + \Sigma_2)/2$ and || signifies the matrix determinant. In the current study, Σ_2 was always a correlation matrix derived from an 18-minute window. Because the time series have equal means as a result of de-meaning during preprocessing, Eq. 1 becomes

$$D_B = \frac{1}{2} \ln\left(\frac{|\Sigma|}{\sqrt{|\Sigma_1||\Sigma_2|}}\right). \tag{2}$$

As Σ_1 and Σ_2 become more similar, D_B approaches zero.

Network Connectivity Rank Reliability. Networks using the 17-Network parcellation were ranked from lowest to highest within-network connectivity. Changes in network ranks across increasing durations of data were evaluated using the Friedman test, a nonparametric repeated measures analysis, and results were corrected for multiple comparisons.

3 Results

3.1 Correlation Matrix Reliability

A steep decrease in D_B , indicating increasingly similar correlation matrices, was observed up to the maximal curvature point observed at durations of 11.21±0.74, 11.97±0.89, and 12.38±1.26 minutes for the HCP Nature, Yeo 17-network, and Yeo-7 network parcellations respectively (Fig. 2). Note that as the number of regions in the

parcellation decreases, the improvement in matrix similarity is less dramatic. D_B then more slowly approached zero. The middle 95% range of D_B values comparing 18minute windows against other 18-minute windows for each parcellation represents the variability of correlation matrices at the maximum duration. The upper limit of this range was taken as a threshold to determine at which duration correlation matrices became comparable to a matrix derived from 18 minutes. The average durations where D_B began to fall within the variability of 18-minute vs 18-minute D_B values were 14.63±1.32, 14.08±1.58, and 13.43±1.97 minutes for the HCP Nature, Yeo 17network, and Yeo-7 network parcellations respectively.

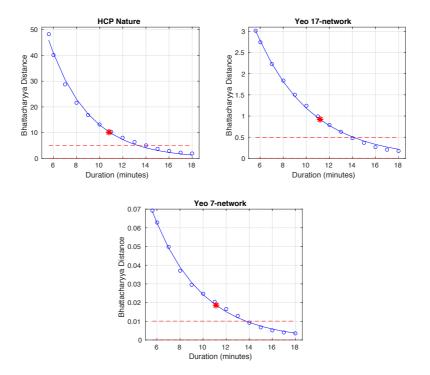


Fig. 2. Mean D_B for each duration compared to 18 minutes of data from three parcellations of a representative subject. Points of maximum curvature (star) and middle 95% ranges of 18min vs 18min D_B values (dotted lines) are shown.

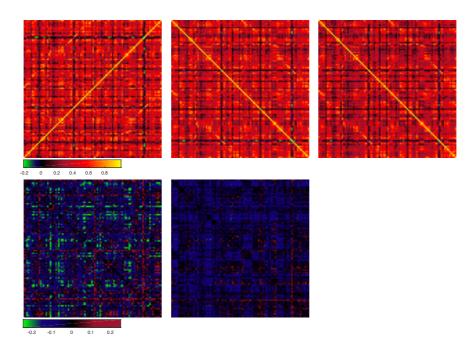


Fig. 3. Top: Representative connectivity matrices derived from 5 minutes (left), 11 minutes (center), and 18 minutes (right) of data using the Yeo 17-network parcellation from a single subject. Bottom: Connectivity differences between 18 and 5 minutes of data (left) and between 18 and 11 minutes of data (right).

3.2 Network Rank Reliability

The Friedman test was used to test for a significant difference in mean within-network connectivity rank across varying durations of data. $93.58\pm6.68\%$ of networks had at least one duration that significantly differed in rank (p<0.029) from of 5.6 to 18 minutes of data. This percentage decreased to $79.14\pm6.64\%$ for rank changes over the range of 10 to 18 minutes, and further to $62.57\pm8.43\%$ from 14 to 18 minutes.

Networks with high within-network connectivity (ranks 16 and 17) were highly reliable even at durations as short as 5-6 minutes (Fig. 4). These ranks primarily corresponded to networks 1 and 2, the peripheral and central visual networks. This agrees with Gonzalez-Castillo, et al. [17] who found the visual network as one of the most temporally stable networks for within-network connections. Many other networks began to display stable rankings at 10-12 minutes of data.

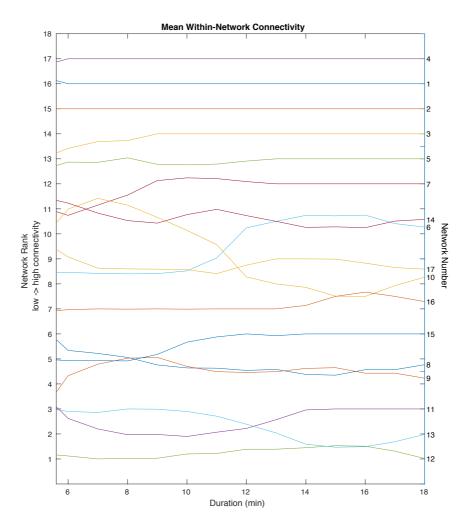


Fig. 4. Mean within-network connectivity ranks from a representative subject using 100 randomly sampled windows of data for each duration.

4 Discussion and Conclusion

Using the Bhattacharyya distance, we found that the reliability of functional correlation matrices derived from rfMRI improves monotonically as more data is acquired. The improvement reaches an inflexion point with windows of data around 11-12 minutes, at which point acquiring more data is less beneficial. Fourteen minutes or more of data was found to produce correlation matrices within the variability of those produced by 18 minutes of data. Since all D_B values were calculated as a comparison to 18-minutes of data, these findings are specific to a study whose gold standard is \sim 20 minutes of rfMRI data. Nonetheless, our method of using the Bhattacharyya distance can be used by other studies to determine how much data is necessary for a session to be usable in further analyses. This is especially valuable for populations that may not tolerate the full scanning protocol of the study.

Our results show that within-network connectivity rank reliability also increases with increasing durations of data, although our statistical tests are not ideally suited for rankings with a zero variance (e.g. a network is **always** ranked the same in all experiments for a particular duration, as network 4 was for all durations except 5.6 minutes (Fig. 4)). Indeed, we found that many networks qualitatively exhibited similar connectivity ranks starting at 10-12 minutes of data, indicating 10-12 minutes of data may be a suitable amount for many network connectivity analyses. Our results also indicate that the duration required is dependent on the network of interest, as some networks, such as the central and peripheral visual network, are very reliable with as little as 5-6 minutes of data.

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