



Functional connectomics of affective and psychotic pathology

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Edited by Marcus E. Raichle, Washington University in St. Louis, St. Louis, MO, and approved March 21, 2019 (received for review December 6, 2018)

Converging evidence indicates that groups of patients with nominally distinct psychiatric diagnoses are not separated by sharp or discontinuous neurobiological boundaries. In healthy populations, individual differences in behavior are reflected in variability across the collective set of functional brain connections (functional connectome). These data suggest that the spectra of transdiagnostic symptom profiles observed in psychiatric patients may map onto detectable patterns of network function. To examine the manner through which neurobiological variation might underlie clinical presentation, we obtained fMRI data from over 1,000 individuals, including 210 diagnosed with a primary psychotic disorder or affective psychosis (bipolar disorder with psychosis and schizophrenia or schizoaffective disorder), 192 presenting with a primary affective disorder without psychosis (unipolar depression, bipolar disorder without psychosis), and 608 demographically matched healthy comparison participants recruited through a large-scale study of brain imaging and genetics. Here, we examine variation in functional connectomes across psychiatric diagnoses, finding striking evidence for disease connectomic “fingerprints” that are commonly disrupted across distinct forms of pathology and appear to scale as a function of illness severity. The presence of affective and psychotic illnesses was associated with graded disruptions in frontoparietal network connectivity (encompassing aspects of dorsolateral prefrontal, dorsomedial prefrontal, lateral parietal, and posterior temporal cortices). Conversely, other properties of network connectivity, including default network integrity, were preferentially disrupted in patients with psychotic illness, but not patients without psychotic symptoms. This work allows us to establish key biological and clinical features of the functional connectomes of severe mental disease.

functional connectome | schizophrenia | major depressive disorder | bipolar disorder | resting-state connectivity

Recent progress in the neurosciences has provided unprecedented opportunities for advancing our understanding of the etiology and pathogenesis of psychiatric illness. At the same time, the gradual reification of diagnostic categories has hampered our ability to take full advantage of these innovations (1–4). To date, the vast majority of research on the biological origins of psychopathology has focused on discrete illness categories, studied in isolation. Although modern psychiatric diagnoses provide advantages to the field in terms of diagnostic reliability, their construct validity and utility for understanding brain circuit dysfunction has been challenged (2, 3). Converging epidemiologic, genetic, and neuroscientific research suggests that populations of psychiatric patients are not separated by clear neurobiological borders between diagnostic categories or across health and disease. There is evidence, for example, of substantial overlap in the genetic factors that increase risk for both affective and psychotic illness (5–7). Consistent with shared heritability, partially overlapping patterns of brain network dysfunction mark a broad range of mental diseases (8–10), indicating that their breakdown can lead to diverse forms of psychopathology. However, despite a flurry of important scientific

advances, we still remain far from a mechanistic understanding of how the functioning of large-scale brain networks might serve to influence suites of behaviors within, or across, psychiatric illnesses.

Identifying signatures of pathology across the functional connectome could provide a framework for researchers to study neurobiological contributions to the onset and maintenance of clinically relevant symptoms, informing the development of novel treatments and future classification schemes. Emerging evidence in healthy populations suggests that individual differences in behavior may be reflected in variability across the collective set of functional brain connections (11–13) (functional connectome) (14). Work from our group and others indicate that the unique connectome architecture of an individual’s brain serves as a stable and reliable “fingerprint” (12, 13, 15–17), likely influenced by genetic variation (18–20). The spectra of symptom profiles observed in patient populations may arise through detectable patterns of network function (1, 21, 22). In particular, the disturbance of individual networks might preferentially contribute to domain-specific (e.g., executive, affective, and

Significance

Historically, most research on the biological origins of psychiatric illness has focused on individual diagnostic categories, studied in isolation. Mounting evidence indicates that nominally distinct psychiatric diagnoses are not separated by clear neurobiological boundaries. Here, we derive functional connectomic signatures in over 1,000 individuals, including patients presenting with different categories of impairment (psychosis), clinical diagnoses, and severity of illness as reflected in treatment seeking. Our analyses reveal features of connectome functioning that are commonly disrupted across distinct forms of pathology, scaling with clinical severity. Conversely, other aspects of network connectivity were preferentially disrupted in patients with psychotic illness. These data have important implications for the establishment of functional connectome fingerprints of severe mental disease.

Author contributions: J.T.B., D.G.D., J.L.R., R.O.B., D.A.P., D.Ö., and A.J.H. designed research; J.T.B., D.G.D., and A.J.H. performed research; J.T.B., L.M.P., and A.J.H. analyzed data; and J.T.B., D.G.D., L.M.P., J.L.R., R.O.B., D.A.P., D.Ö., and A.J.H. wrote the paper.

Conflict of interest statement: Over the past 3 years, D.A.P. has received consulting fees from Akili Interactive Labs, BlackThorn Therapeutics, Boehringer Ingelheim, Compass, Posit Science, and Takeda Pharmaceuticals and honoraria from Alkermes for activities unrelated to the current review. J.T.B. has received consulting fees from Pear Therapeutics and Niraxx Therapeutics. J.L.R. has received investigator-initiated funding from PamLab. D.Ö. served on an Advisory Board for Neurocrine Inc. in December 2016, unrelated to the current work. No funding from these entities was used to support the current work, and all views expressed are solely those of the authors.

This article is a PNAS Direct Submission.

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This article contains supporting information online at www.pnas.org/lookup/suppl/doi:10.1073/pnas.1820780116/-DCSupplemental.

Published online April 15, 2019.

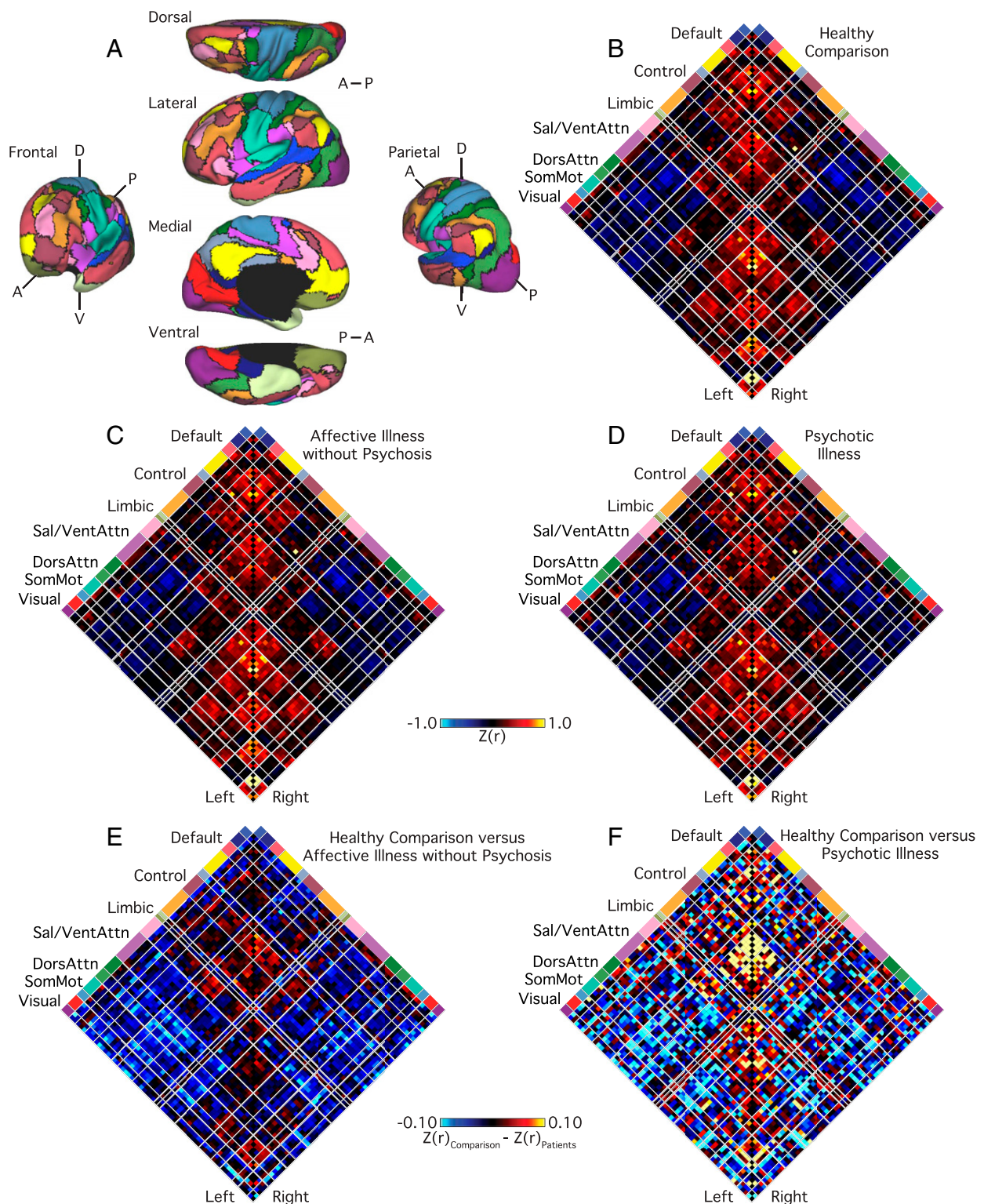


Fig. 1. Cortical network connectivity in patients and healthy comparison participants. (A) The functional network organization of the human cerebral cortex revealed through intrinsic functional connectivity. Colors reflect regions estimated to be within the same network. Regions determined based on the 17-network solution from Yeo et al. (34). The approach groups similar correlation profiles based on a winner-take-all solution, with every surface vertex assigned to its best-fitting network. The 2D grids (B–D) display the complete coupling architecture of the cerebral cortex measured at rest for (B) the healthy comparison participants, (C) patients with affective illnesses without psychosis, and (D) patients with psychotic illnesses. Values reflect z-transformed Pearson correlations between every region and every other region, after accounting for the effects of coil, scanner, console software version, age, sex, race, ethnicity, and handedness. Within-network correlations fall along the center diagonal. Between-network correlations are plotted away from the diagonal and reveal both positive (red) and negative (blue) correlations. (E and F) The 61×61 grids show the differences in resting BOLD correlation between controls and (E) patients with affective illnesses without psychosis, as well as (F) patients with psychotic illnesses, for each intrahemispheric regional pair. Differences were obtained by an analysis of variance of z-transformed Pearson correlation values, adjusting for nuisance variables. White lines represent network boundaries. DorsAttn, dorsal attention; Left, left hemisphere; Right, right hemisphere; Sal, salience; SomMot, somatomotor; and VentAttn, ventral attention.

abnormal connectivity across the connectome, with the disturbance of individual systems preferentially contributing to certain symptom domains that can present in a disorder-general manner.

Our findings of transdiagnostic disruptions in frontoparietal network connectivity are consistent with prior work in both schizophrenia (39–41), unipolar depression (27, 42, 43), and bipolar disorder (44, 45), where there is converging evidence for abnormalities in cognitive control and context processing (8, 21). By studying multiple patient populations simultaneously, without prejudice toward ascertainment or diagnostic label, our findings allow for the simultaneous characterization and comparison of psychiatric connectomes across both affective and psychotic illnesses. Furthermore, while much of the prior work in this domain has focused on circumscribed profiles of dysfunction in either dorsolateral prefrontal or anterior cingulate cortices, here we provide evidence indicating broad frontoparietal network impairments that span aspects of frontal, parietal, temporal, and medial prefrontal components of this network.

Although these data support the view that the frontoparietal network may underlie a diverse set of cognitive processes impaired in multiple disorders, one outstanding question centers on the extent to which impaired frontoparietal connectivity in individual patients may reflect a primary factor associated with disease onset and maintenance, or a secondary consequence of illness (8). In depression, for example, deficits on neuropsychological measures of executive functioning track both the severity of current symptoms in patients, as well as the use of psychotropic medications (35). Suggesting the likely presence of preillness shifts in network connectivity, prefrontal dysfunctions related to context processing have been identified in never-medicated patients with schizophrenia early in the course of the illness (39). Highlighting a degree of symptom specificity, in a study of psychotic patients with and without cognitive dysfunction, we recently reported that distinct frontoparietal subnetworks may link to cognitive capacity (i.e., control A subnetwork) and psychiatric symptoms (i.e., control B subnetwork), respectively (46). Across development the human brain experiences distinct functional changes (47) as network modules become more segregated with age (48). Impairments in this process of network differentiation, for example between default and executive (frontoparietal) networks, are linked to dimensions of psychopathology that cross traditional diagnostic boundaries (10). Despite the importance of distinguishing network-level risk factors from markers of current symptom severity, the extent to which connectome functioning parallels clinical trajectories across the lifespan remains to be determined.

A considerable body of evidence has accumulated over recent years suggesting that the presence of altered default network functioning may mark psychotic illness (49). Encompassing aspects of ventral and dorsal medial prefrontal, posterior/retrosplenial, and inferior parietal cortices, the default network is hypothesized to underpin self-referential processing and principal aspects of mental simulation (50). Core symptoms of psychotic illness arise from misattributions of thought and a blurring of the boundaries separating internal cognition from the external world (50, 51). Consistent with our reported analyses, these converging lines of evidence suggest an association linking impaired default network functioning and the occurrence of psychotic symptoms (e.g., hallucinations, delusions, and thought distortions).

Critically however, the present results should not be taken to suggest that default network disruption is only present in patients with psychotic illness. Indeed, in line with prior reports (52), a muted decrease in default network functioning was observed in patients with bipolar disorder without psychosis. Rather, our data support the view that default network functions may underlie a set of cognitive processes impaired in multiple disorders (49), with variability in network-level functioning linking with corresponding network-associated symptom expression (21).

Patients with unipolar depression, for example, have been found to display aberrant intrinsic connectivity (53) as well as heightened stimulus-induced activity in aspects of the default network while viewing and reappraising negative images (54). These data are consistent with our observation of increased default D subnetwork connectivity in nontreatment-seeking individuals with depression, a functional profile that was absent in treatment-seeking patients recruited from clinical units. Given this observed variability within diagnoses, future high-throughput data-collection efforts will be necessary to establish the manner in which individual specific connectome architecture might serve as a dimensional fingerprint of human behavior, predicting symptom profiles in patient populations with varying degrees of clinical severity (35).

Analyses that link functional connectomes to individual differences in behavior, symptom profiles, and severity of illness represent a tremendous opportunity for the field. The present analyses suggest that the severity of psychotic and manic symptoms may emerge, at least in part, through common profiles of functional variability. However, it is important to note that although these clinical phenotypes may arise through a shared network architecture, they are not interchangeable. Rather, our results likely reflect a general role for aspects of frontoparietal and default networks in the regulation of cognitive processes that broadly underlie affective and psychotic illness. While it may not be feasible to identify isolated features of brain biology that cleanly distinguish populations of patients with psychiatric illness, multivariate fingerprints of pathology may eventually emerge. To establish such points of separation, our data collection and analytic efforts need to incorporate dimensional measures of clinical severity across a broad range of patient populations, recruited from diverse clinical settings at varying phases of illness. To generate such high-dimensional datasets, we will need to reassess our current scientific approach, extending beyond conventional clinic- or research laboratory-specific collection efforts. Our present analyses reflect the combined efforts of multiple research groups collaborating to collect data with a harmonized acquisition sequence (30). This cross-laboratory collaborative effort allowed us to partially disentangle the relations between clinical diagnoses and degree of treatment seeking.

Readers should note that there are limitations to the conclusions that can be drawn from the present analyses. Given the cross-laboratory collaborative nature of the present work, a consistent self-report or task battery was not available for analysis across the participants. Accordingly, we are unable to make claims regarding associations that may link network function with the presence, absence, or severity of specific dimensional symptom profiles. To make progress in this domain, our clinical recruitment strategies and analytic efforts will need to coordinate across research laboratories to standardize imaging acquisition protocols, as well as dense demographic, symptom, and behavioral batteries (1). Additionally, while we can compare and contrast treatment-seeking and nontreatment-seeking individuals with unipolar depression, we are unable to account for other factors that may contribute to access to care and utilization. For example, the nontreatment-seeking individuals with depression were not currently taking psychiatric medications. This is in contrast to individuals already in treatment who were prescribed varying forms of psychiatric medication. Moreover, lack of insight (55), as well as internalized and treatment stigma, can associate with reduced help seeking in some patient populations (56). Consequently, despite evaluating over 1,000 individuals, we are limited in the conclusions we can draw when comparing and contrasting the connectomic profiles observed in nontreatment-seeking individuals recruited from the community versus populations of patients recruited from partial and inpatient hospitalization programs. Smaller-scale longitudinal studies suggest the presence of abnormal functional connectivity

in individuals presenting to the emergency room seeking care for schizophrenia or related diagnoses, even before starting psychiatric medication (57). Moreover, connectomic changes are apparent in early phases of antipsychotic treatment (58). This literature suggests that differences in symptom severity, rather than medication per se, may underlie the extent and degree of changes observed in our present analyses. However, future longitudinal research designs will be critical for fully disentangling the effects of treatment, fluctuating symptom-severity, and illness course on brain function.

The unprecedented growth of big data in neuroscience provides opportunities for researchers seeking to understand how brain functions influence suites of behaviors and associated illness risk. In the present analyses we make use of a large sample of individuals with imaging data, spanning domains of psychopathology, levels of acuity, and engagement with care. This heterogeneous sample of participants represented a broad range of symptom profiles and illness severity, including individuals with self-reported mental health, nontreatment-seeking forms of depression, and treatment-seeking forms of unipolar depression, bipolar disorder, and severe psychotic illness. Our analyses revealed aspects of the frontoparietal control network that are commonly disrupted across diagnostically distinct forms of severe pathology, whether psychotic or nonpsychotic affective in nature. In addition, we established both shared and unique functional alterations in affective and psychotic illnesses. For example, a preferential reduction in default network integrity was evident in patients with psychotic illness, but absent in affective illnesses without psychosis. These analyses highlight the potential to discover individualized network profiles that are predictive of symptom-relevant cognitive domains, both within and across diagnostic boundaries, as exemplified in the Bipolar-Schizophrenia Network on Intermediate Phenotypes (B-SNIP) effort (59) and our own ongoing work. In conclusion, this study provides a comprehensive characterization of connectomic dysfunction in a range of psychopathological conditions that matches well with the core deficits observed in these populations. These data have important implications for the future creation of connectome-based models that predict behavior, an approach with the potential to account for symptom comorbidity while simultaneously explaining the biological process that give rise to the diversity of clinical presentations.

Methods

Between November 2008 and June 2017, fMRI data were collected from a total of 1,010 individuals, including 210 diagnosed with a primary psychotic disorder (137 meeting criteria for schizophrenia or schizoaffective disorder, 73 with bipolar disorder with psychosis), 192 presenting with a primary affective disorder without psychosis (26 with bipolar disorder without psychosis, 57 treatment-seeking individuals with unipolar depression, 109 nontreatment-seeking individuals with unipolar depression), and 608 demographically matched healthy comparison participants recruited through an ongoing, large-scale study of brain imaging and genetics (30). Diagnosis was determined using the Structured Clinical Interview for the DSM-IV (60). Details regarding participant recruitment and characterization, as well as the demographic and clinical characteristics of the patient and matched healthy comparison samples, are available in *SI Appendix, Table S1*. In brief, patients were recruited from clinical services at MGH or McLean Hospital through the procedures detailed in Baker et al. (23). Nontreatment-seeking individuals who met diagnostic criteria for unipolar depression were recruited from the surrounding Boston area using the procedures detailed in Dillon et al. (61).

Healthy comparison participants were selected from an existing database of adults (aged 18–83 y) (30), scanned previously using identical pulse sequences on identical scanners, and selected to match patients on the basis of age, gender, race, handedness, as well as a mean slice-based signal-to-noise ratio (SNR) derived from the participant's blood oxygenation level-dependent (BOLD) T2* image series. In this context, SNR is calculated as the mean/SD of the mean slice intensity time series. Using this strategy, we were able to ensure statistically matched distributions for both demographic variables and comparable data quality (as well as head movement metrics).

The reported experiments were approved by the Partners HealthCare Institutional Review Board and the Harvard University Committee on the Use of Human Subjects in Research McLean Hospital Institutional Review Board, and all participants gave written informed consent before participating in the study.

MRI Data Acquisition. Imaging data were collected on 3T Tim Trio scanners (Siemens) using either 12- or 32-channel phased-array head coils at Harvard University, MGH, or McLean Hospital as detailed in Holmes et al. (30). Briefly, structural data included a high-resolution, multiecho T1-weighted magnetization-prepared gradient-echo image [144 slices, repetition time (TR) = 2,200 ms, inversion time (TI) = 1,100 ms, echo time (TE) = 1.54 ms for image 1 to 7.01 ms for image 4, flip angle = 7°, voxels = 1.2 mm³, field-of-view (FOV) = 230]. Functional data were acquired using a gradient-echo echoplanar imaging sequence (47 axial slices, interleaved with no gap), 124 time points (TR = 3000 ms, TE = 30 ms, flip angle = 85°, voxels = 3 mm³, FOV = 216). Participants were instructed to remain still and keep their eyes open, while blinking normally. Although no fixation image was used, participants with psychotic illness were monitored via eye-tracking video to ensure compliance during functional scans. Software upgrades (VB13, VB15, VB17) occurred during data collection. All results are reported after partialing out variance associated with coil, scanner (Harvard Bay 1, McLean Bay 1, MGH Bay 4, MGH Bay 8, and so forth), and software upgrade, as well as age, sex, handedness, race, and ethnicity. All treatment-seeking patient samples were collected on a 12-channel coil. In the healthy comparison, participants and nontreatment-seeking individuals with unipolar depression 78.5 and 36.7% of the data were collected on a 12-channel coil, respectively. All reported analyses are consistent when separately considering only 12-channel and 32-channel coil data. The patient and healthy comparison samples did not differ in mean slice-based signal-to-noise [all patients: 172.4 ± 66.8; healthy comparison: 175.3 ± 51.2; $F_{(1, 1,008)} = 0.61, P = 0.43$]. Patients displayed a significantly greater number of micromovements (translations > 0.1 mm) during data collection [all patients: 25.5 ± 27.2; healthy comparison: 20.3 ± 24.7; $F_{(1, 1,008)} = 9.89, P \leq 0.005$]. The reported group-level effects are consistent when incorporating mean slice-based SNR and micromovement counts as model covariates.

Preprocessing. Data were analyzed with a series of steps common to intrinsic connectivity analyses (31–33) and further elaborated in Holmes et al. (30) and Yeo et al. (34). Preprocessing included discarding the first four volumes of each run to allow for T1-equilibration effects, compensating for slice acquisition-dependent time shifts per volume, and correcting for head motion using rigid body translation and rotation. Additional steps involved the removal of constant offset and linear trends over each run and the use of a temporal filter to retain frequencies below 0.08 Hz. Sources of spurious variance, along with their temporal derivatives, were removed through linear regression. These included six parameters obtained by correction for rigid-body head motion, the signal averaged over the whole brain, the signal averaged over the ventricles, and the signal averaged over the deep cerebral white matter. Functional data were first aligned to the structural image using the FreeSurfer software package, smoothed using a 6-mm kernel applied in surface space, and down-sampled to a 4-mm mesh Yeo et al. (34).

Functional Parcellation. Cortical functional coupling matrices were computed for each participant, across all available regions within the 17 network functional parcellation of Yeo et al. (34) (Fig. 1A). This parcellation consisted of 122 cortical regions composed of 61 roughly symmetric territories in the left and right hemispheres (23). Correlation matrices were constructed to include all regional pairs arranged by network membership. Pearson correlation coefficients were computed between each regional fMRI time course, averaged across all vertices within the region, and the mean fMRI time course for every other region (Fig. 1B–D). Correlation values were z-transformed to increase normality of the correlation distribution and compared across groups using an ANOVA after linear regression of nuisance variables. Reported tests survived correction for multiple comparisons using a family-wise error rate (Bonferroni procedure) of $P \leq 0.05$ or FDR of $q \leq 0.05$. Readers should note that caution is warranted when interpreting group differences in within-network connectivity for subnetworks with limited numbers of parcels (e.g., frontoparietal control C and default D).

ACKNOWLEDGMENTS. We thank the Simons Foundation and the Howard Hughes Medical for their invaluable contributions on Randy L. Buckner's work on the GSP; and Kevin Dowling for facilitating access to clinical rating scale data. This work was supported by the Brain & Behavior Research Foundation (A.J.H., J.L.R., and D.G.D.); the Taplin Family Foundation (D.Ö.); PamLab (J.L.R.); and National Institute of Mental Health Grants R01MH101425 (to J.L.R.),

K23MH104515 (to J.T.B.), F32MH081394 and R00MH094438 (to D.G.D.), R37MH068376 and 1R01MH101521 (to D.A.P.), K23MH100623 (to R.O.B.), K23MH079982-01A1 and R01MH094594 (to D.Ö.), and K01MH099232 (to A.J.H.). Analyses were made possible by the resources provided through Shared Instrumentation Grants 1S1ORR023043 and 1S1ORR023401. Data were provided in part by the Brain Genomics Superstruct Project of Harvard University and Massachusetts General Hospital (Principal Investigators: Randy L. Buckner, Jordan W. Smoller, and J.L.R.) with support from the

Center for Brain Science Neuroinformatics Research Group; the Athinoula A. Martinos Center for Biomedical Imaging; and the Center for Genomic Medicine. The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health. The funding organizations had no role in the design and conduct of the study; collection, management, analysis, and interpretation of the data; preparation, review, or approval of the manuscript; and decision to submit the manuscript for publication.

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