Baseline brain structural and functional predictors of clinical outcome in the early course of schizophrenia

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Abstract

Although schizophrenia is considered a brain disorder, the role of brain organization for symptomatic improvement remains inadequately defined. We investigated the relationship between baseline brain morphology, resting-state network connectivity and clinical response after 24-weeks of antipsychotic treatment in patients with schizophrenia (n = 95) using integrated multivariate analyses. There was no significant association between clinical response and measures of cortical thickness (r = 0.37, p = 0.98) and subcortical volume (r = 0.56, p = 0.15). By contrast, we identified a strong mode of covariation linking functional network connectivity to clinical response (r = 0.70; p = 0.04), and particularly to improvement in positive (weight = 0.62) and anxious/depressive symptoms (weight = 0.49). Higher internal cohesiveness of the default mode network was the single most important positive predictor. Key negative predictors involved the functional cohesiveness of central executive subnetworks anchored in the frontoparietal cortices and subcortical regions (including the thalamus and striatum) and the inter-network integration between the default mode and sensorimotor networks. The present findings establish links between clinical response and the functional organization of brain networks involved both in perception and in spontaneous and goal-directed cognition, thereby advancing our understanding of the pathophysiology of schizophrenia.

Introduction

Schizophrenia features amongst the leading causes of disability worldwide [1] because it is associated with chronicity [2, 3], psychosocial disadvantage [4, 5] and reduced life expectancy [6, 7]. The five years following the first psychotic episode are considered a critical period as the long-term pattern of the disease course is typically established during this time [8, 9]. Prior research on the prognostic significance of clinical and demographic features has highlighted the importance of symptomatic improvement in response to antipsychotic treatment both in clinical trials and in observational studies [10–13]. Although schizophrenia is reliably associated with alterations in brain structure [14–16] and functional connectivity [14, 17, 18], the relevance of brain organization for symptomatic response is poorly understood. Magnetic resonance imaging (MRI) studies have produced encouraging results in terms of identifying prognostic biomarkers in other brain disorders, such as epilepsy [19] and Alzheimer’s disease [20] and this evidence motivates research efforts to identify brain imaging predictors of treatment response in schizophrenia.

The neuroimaging literature thus far has been encouraging but inconclusive. Despite negative results [21–24], several structural (sMRI) studies have linked poor treatment response to reductions in gyriﬁcation [25] in global [26], prefrontal [27] and temporal [28] cortical gray matter, and in parahippocampal [29, 30] and striatal volumes [31]. The functional (fMRI) literature is silent about the role of large-scale brain networks since the few available studies have each examined the connectivity of specific regions of interest. Speciﬁcally, Nejad and colleagues [32] reported that baseline frontoparietal connectivity during a working-memory task distinguished patients with persistent and non-persistent negative symptoms after 6 months of antipsychotic treatment. Kraguljac and colleagues [33] found...

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that higher baseline hippocampal connectivity with the dorsal anterior cingulate, caudate, and auditory cortex coupled with lower connectivity with the lingual gyrus predicted response after 6 weeks of antipsychotic treatment. Finally, Sarpal and colleagues [34] showed that antipsychotic response at 12 weeks was associated with lower striatal connectivity with the anterior cingulate and medial prefrontal cortex and higher striatal connectivity with posterior brain regions.

There are at least four challenges to be addressed in moving the field forward. First, previous studies divided patients into subgroups using various definitions of response derived from consensus statements (e.g., Remission in Schizophrenia Working Group [35]). Here we adopt a dimensional rather than a dichotomous approach because regardless of the specific definition, these categorical subgroups do not have meaningful differences in symptom severity and do not resonate with the experience of patients and carers [36]. Second, we expand previous literature by considering the entire spectrum of psychopathology in schizophrenia, which includes depressive/anxious symptoms and agitation/disorganization [37] in addition to positive and negative symptoms. This dimensional symptom structure is present at the first psychotic episode and is longitudinally stable [37] (further details in supplemental information (SI)). Therefore, inclusion of all four-symptom dimensions should be integral to any attempts to relate neuroimaging measures to clinical improvement. Third, our analyses also consider body mass index (BMI) and IQ, that represent robust indices of cardiometabolic health [38] and cognition [39, 40], respectively, are strongly associated with brain organization [41] and with symptom severity [42, 43]. Inclusion of these key variables when modeling the relationship of neuroimaging features to clinical response increases confidence that the findings reflect schizophrenia–related neurobiological processes. Fourth, we extend the scope of inquiry beyond testing regionally specific hypotheses which is critical for the detection of functional connectivity predictors of clinical response that are not a priori predicted.

Here, we acquired sMRI and resting-state functional (rs-fMRI) neuroimaging data from patients at the early stages of schizophrenia prior to initiation of antipsychotic treatment. Clinical response was assessed in terms of change in positive, negative, anxious/depressive and agitation/disorganization symptoms 24 weeks later. IQ and BMI were also assessed as part of the clinical dataset. We used sparse Canonical Correlation Analyses (sCCA) [44] to investigate the underlying relationship between the neuroimaging and clinical datasets as this multivariate method is ideally suited for predicting one dataset from the other while accounting for correlations between variables [45, 46]. Our aim of this data-driven approach was to identify previously undetected associations between whole-brain structural and functional neuroimaging features and clinical response following treatment. When considering preexisting evidence our working hypotheses were that measures of cortical thickness and of functional integration between cortical and subcortical regions would emerge as the most significant predictors of symptomatic change.

**Materials and methods**

**Study design**

Patients within the first 5 years of being diagnosed with schizophrenia based on the criteria of the Diagnostic and Statistical Manual of Mental disorders, Fifth Edition (DSM-5) [47] were recruited via clinician referrals from the psychiatric services of the Mount Sinai Health System at the Icahn School of Medicine at Mount Sinai (ISMMS), New York. Clinicians were asked to refer patients who were willing to re-start antipsychotic treatment following a minimum 2-week medication free period based on patients’ self-report. Patients were screened to exclude individuals with IQ <70 and those with treatment refractory illness, history of head trauma or loss of consciousness, pervasive developmental disorder, lifetime history of DSM-5 substance use disorder, medical or neurological disorders that could cause psychiatric symptoms and contraindications to MRI. Eligible patients were enrolled prior to restarting antipsychotic treatment. Brain MRI scans were acquired as soon as patients could tolerate this examination to ensure minimal exposure to antipsychotic medications. Patients were evaluated clinically on the day of scanning and then 24 weeks later. The study was approved by the institutional review board of the ISMMS; all participants provided written informed consent prior to study enrolment.

**Clinical assessment**

Enrollment only: The patients’ diagnostic status was confirmed with the Structured Clinical Interview for DSM-5 [48]. Current IQ was obtained with the Wechsler Abbreviated Scale of Intelligence (WASI-II) [49]. IQ was used as an index of overall cognitive ability because it correlates with domain-specific cognitive tests [39] and accounts for most of the variability in cognitive function [40]. In the context of schizophrenia, IQ is also associated with variability in the clinical profile and brain structure of patients [41, 50].

Day of scanning and at 24-week follow-up: Psychopathology was assessed using the 24-item Brief Psychiatric Rating Scale (BPRS) [51] because it can reliably assess positive and negative symptoms, agitation/disorganization...
and anxiety/depression [52] (Supplementary Table 1). Medication type and dose were recorded in all patients and supplemented with information from their medical records. The daily antipsychotic dose was converted to chlorpromazine equivalents (CPZE) [53]. The BMI was included as a general measure of cardiometabolic health [38] with known associations with symptom severity [42] and brain structure in patients with psychosis [41].

**Treatment and evaluation of clinical outcome**

The choice and dosing of the antipsychotic agent was based on the judgment of the patients’ treating psychiatrists since differences in the effectiveness of individual antipsychotics are small [54, 55] and appropriate treatment decisions crucially depend on patients’ history, preference and circumstances. During the inter-assessment interval, patients received routine clinical care which consisted of outpatient appointments with their treating psychiatrist. These appointments involved informal supportive therapy and monitoring of mental state, adverse events and adherence. Patients were seen weekly for the first three weeks post-treatment initiation. Thereafter, planned appointments occurred depending on individual patient’s needs and preferences. Adherence was monitored via self-report during appointments and through the Mount Sinai electronic prescription records (also in SI). None received psychological or electroconvulsive therapy. Additional support was provided by social workers in connection to disability benefits and accommodation. The primary outcome was the change in the BPRS subscale scores for positive, negative, anxious/depressive and agitation/disorganization symptoms between initial (T1) and 24-week follow-up assessment (T2) calculated as \( \frac{\text{T}_1 - \text{T}_2}{\text{T}_1} \); positive and negative values respectively represent symptomatic improvement or worsening. We adopted a dimensional approach to symptomatic change rather than dichotomous remission/non-remission categorizations for the following reasons: (i) although there are several criteria for remission there is still no consensus on the cut-offs that define remission in schizophrenia; (ii) it is questionable whether clinically defined remitted and non-remitted patients (regardless of the criteria applied) represent biologically distinct groups; (iii) although there is some evidence that treatment resistant patients may have a distinct neurobiological profile such patients were excluded here by design. In support of these statements we conducted supplemental analyses detailed in SI.

**Neuroimaging**

MRI acquisition and processing: We acquired high-resolution sMRI and rs-fMRI data on the same 3 T Skyra scanner (Siemens Medical Systems, Erlangen, Germany) using identical protocols for all participants. Comprehensive quality control (including head motion) was carried out on all neuroimaging datasets using previously published quality control procedures. Details of the image acquisition, preprocessing and quality assurance are provided in the SI.

Cortical reconstruction and volumetric segmentation: In each individual sMRI dataset, we used FreeSurfer image analysis suite (v.5.3.0) (http://surfer.nmr.mgh.harvard.edu/) to derive 64 cortical thickness measures and 18 subcortical volumetric measures from the Desikan atlas [56][details in SI and Supplementary Figure 1].

Computation of resting-state connectivity: We used previously established and reproducible templates [57, 58] based data from 496 participants of the Human Connectome Project (www.humanconnectomemproject.com) to partition the functional connectome into the 5 most replicated resting-state networks (RSNs) and their subdivisions: default mode (DMN; 5 subnetworks), salience (SAL), central executive (CE; 3 subnetworks), sensorimotor (SMN; 2 subnetworks) and visual (VIS; 2 subnetworks) (Fig. 1, Supplementary Table 2). This partition was based on averaged regional blood oxygen-level-dependent (BOLD) time series from each of 638 brain regions chosen to maximize the likelihood of detecting subtle effects at the level of subnetworks as per Zalesky et al [59], and Crossley et al. [60]. In each participant, Fisher Z-transformed Pearson’s correlation coefficients were computed to calculate functional connectivity within (i.e., cohesiveness) and between (i.e., integration) subnetworks. Cohesiveness was computed as the average correlation of each voxel’s BOLD signal time series with every other gray matter voxel within the subnetwork. Integration was computed as the correlation between the average time-series of each pair of subnetworks. These computations yielded 91 connectivity measures.

**Statistical analyses**

We used paired t-tests to examine differences in symptom severity, BMI and antipsychotic dose between initial and 24-week assessments. We used univariate correlations and curve estimation to determine the zero-order associations between clinical and neuroimaging features (Supplementary Figure 3). We conducted sCCA to determine the covariation patterns between clinical features (clinical dataset) with sMRI (sMRI dataset) and with rs-fMRI (rs-fMRI dataset) neuroimaging measures. The clinical dataset comprised sex, IQ and change in each of the BPRS subscale scores and in BMI and antipsychotic dose (Supplementary Table 1). The sCCA can be considered as a generalization of multiple linear regression and aims to identify those linear combinations (pairs of canonical variates or otherwise modes) of variables in each dataset that best express the maximal correlation (i.e., canonical correlation) between the two variables.
datasets. The weight of association between a variable of one dataset to the variate of the other dataset can be conceptually construed as an extension of a correlation coefficient (additional details in SI). All sCCAs were implemented in MatlabR2015b using an in-house script according to our previously published work [41, 61]. Age and the interval between T1 and T2 were then regressed out of the change scores. Continuous variables were z-standardized prior to being entered into the sCCA. The sMRI dataset comprised 82 variables corresponding to measures of cortical thickness and subcortical volume (Supplementary Figure 1). The rs-fMRI dataset comprised 91 variables corresponding to measures of RSN cohesiveness and integration. Age was regressed out of the sMRI and fMRI variables before they were z-standardized and entered in the sCCA.

For each sCCA, we (a) computed the sparse parameters for range of candidate values from $0.1 \times \sqrt{p}$ (high sparsity) to $1 \times \sqrt{p}$ (low sparsity) at 0.1 increments (where $p$ is the number of features in that view of the data) and fitted the resulting models; (b) selected the optimal sparse criteria combination based on the parameters that corresponded to the values of the model that maximized the sCCA correlation value; (c) determined the optimal sCCA model and established its significance at $p < 0.05$ using permutations ($n = 10,000$). The $p$-value was defined as the number of permutations that resulted in a higher correlation than the original data divided by the total number of permutations. Thus the $p$-value is explicitly corrected for multiple testing as it is compared against the null distribution of maximal correlation values across all estimated sCCAs. For each permutation, we tested all sparsity criteria combinations as for the original data and then extracted the sCCA correlation with the highest coefficient among the tested options, independently of whether this combination was the same as in the original data. In this way we ensured that we did not underestimate the chance of a permutation achieving the same or higher value than the original data. If the sCCA was significant, we reported the canonical correlation weight of each variable.

**Reliability analyses**

We confirmed that the results reported were not influenced by the choice of functional partition, by repeating the analyses using measures of functional cohesiveness and integration extracted from other publicly available RSN masks. These analyses are detailed in the SI.
Supplemental categorical analyses

Although we adopted a dimensional approach for our core analyses, we also tested the usefulness of categorical definitions of remission as we detail in SI. Specifically we investigated (a) whether we could identify differences in neuroimaging features when patients were categorized into remitted and non-remitted using the three most widely used criteria, and (b) whether we could identify clusters of patients based on their imaging features that differed in terms of their remission status.

Results

Clinical features

Ninety-five patients were initially enrolled and 19 were excluded from the final analyses (7 because poor quality scans and 12 because they declined the follow-up interview). The demographic and clinical characteristics of the final sample are shown in Table 1. The majority of the patients were started on second generation antipsychotics [most commonly risperidone/paliperidone (64%) and olanzapine (26%)] while 12 patients (15.8%) were prescribed first generation agents [most commonly haloperidol (58%) and fluphenazine (41%)]. Overall, there was a significant reduction in all symptom dimensions at the follow-up assessment (Table 1). There were no significant differences in any clinical or demographic information (i.e., age, disease duration, age at onset, BMI, medication change, IQ, change in symptoms) between men and women (all \( p > 0.07 \) at an uncorrected threshold). Although we adopted a dimensional approach, here we also report rates of remission based on the Remission in Schizophrenia Working Group (RSWG) criteria [35] to allow comparisons of the clinical data with other studies that have used dichotomous categorizations. Consistent with the rates reported in prior literature [62], 50 (65.8%) patients were remitted based on RSWG criteria; as expected there was significant overlap in the distribution of symptom scores between remitted and non-remitted patients (Supplementary Figure 2).

Brain structural predictors of clinical response

sCCA models that considered all the structural data simultaneously (\( r = 0.54, p = 0.3 \)) and models that were restricted to either subcortical (\( r = 0.56, p = 0.15 \)) or cortical measures (\( r = 0.37, p = 0.98 \)) were not significant. This pattern was also seen at the level of univariate correlations between clinical and sMRI measures were small (\( r \) values)

![Figure 2](image-url)
Brain functional connectivity predictors of clinical response

Only the first mode of the sCCA of the clinical and rs-fMRI datasets was significant ($r = 0.70$, $p = 0.04$) (Fig. 2a). The weight of the association with imaging variate was highest for positive symptoms (weight = 0.62) followed by anxiety/depression (weight = 0.49), negative symptoms (weight = 0.33) and agitation/disorganization (weight = 0.32) (Fig. 2b). The weights for BMI or antipsychotic dose were zero. The association between clinical variables and the rs-fMRI variate was higher for women (Supplementary Figure 4); however, removal of sex did not change the

Fig. 3 Brain functional predictors of symptomatic response in schizophrenia. a Links depict the canonical weights of the association between the clinical variate and the connectivity of the subnetworks depicted; Red links represent positive predictors, Blue links represent negative predictors. b and c provide illustrative examples of the associations between functional integration and improvement in positive symptoms. Improvement in positive symptoms was chosen because it had the highest canonical weight to the rs-fMRI variate. DMN1 = medial temporal default mode subnetwork; DMN3 = anterior precuneus/posterior cingulate cortex; DMN4 = dorsal posterior precuneus; SMN2 = sensorimotor network, dorsal subnetwork; rs-fMRI resting state functional magnetic resonance imaging; details of the spatial distribution of the networks are provided in Fig. 1 and Supplementary Table 2.
overall results. We focus on variables with at least moderate absolute weights (≥0.20, Fig. 3) and we provide details of all the weights of association between the rs-fMRI measures and the clinical variate in Supplementary Table 3. Positive weights indicate that the corresponding variables are positive predictors while negative weights denote negative predictors. Higher functional integration of DMN subnetworks (DMN1 with DMN4 and DMN5), and particularly higher connectivity between medial temporal regions and the dorsal precuneus (PCu), emerged as the main positive predictor of response (Fig. 3). The negative predictors comprised features of cross-network integration and within-network cohesiveness. Key negative predictors consisted of the connectivity between the DMN3 and SMN2 (i.e., functional connectivity between the PCu/posterior cingulate cortex and sensorimotor cortices), between the DMN5 and SAL (i.e., functional connectivity between the PCu and regions of the salience network that mainly involve the insula and anterior cingulate cortex) and between the CEN1 and the VIS1 (i.e., functional connectivity between the lateral frontoparietal and lateral visual cortices) (Fig. 3). Key functional cohesiveness predictors implicated the CEN1 (comprising frontoparietal regions) and CEN3 (comprising the thalamus, striatum and posterior hippocampus).

Supplemental results

As detailed in SI, adopting a categorical approach to data analysis did not yield statistically significant results suggesting that remission and non-remission are more likely to represent a continuum rather than biological distinct states.

Discussion

We investigated the role of baseline sMRI and rs-fMRI measures as predictors of symptomatic improvement in response to antipsychotic treatment in the early course of schizophrenia. The present results establish the importance of system-level functional brain organization and clinical response. Specifically, the cohesiveness of the DMN and CEN and their cross-network integration with the salience, somatosensory and visual networks emerged as key predictors of short-term symptomatic change in schizophrenia. Moreover, our supplemental analyses argue against the notion that remission can be predicted by distinct functional connectivity features and suggest that short-term symptomatic change in schizophrenia is dimensionally linked to alterations in functional connectivity.

These findings can be best understood within the context of the developmental trajectories and organizational principles of the human connectome. At the highest level, the functional connectome of healthy adults is intrinsically organized into two anticorrelated large-scale neural systems: the DMN which comprises midline brain regions associated with unconstrained mental activity and a neural system preferentially engaged by goal-directed cognition and sensorimotor processes [58, 63–66]. The latter is functionally distinguished into higher-order networks, notably the CEN (concerned with maintenance and manipulation of goal-directed mental operations) and the SAL (involved in switching activity between the DMN to CEN) and in lower-order sensory and motor networks [63, 64]. Transition from childhood to adulthood typically is associated with changes in the intrinsic functional connectivity of these networks. For the DMN this entails greater intra-network cohesiveness and greater inter-network integration with the higher-order networks [58, 67]. Gu et al. [58] proposed that this pattern of connectivity allows the DMN to function as cohesive driver of intrinsic brain function that supports a range of mental operations relating mostly to self-referential processes and spontaneous thought [66, 68]. The development of critical DMN connectivity may be disrupted in schizophrenia based on previous reports of reduced functional cohesiveness between DMN nodes and particularly the hippocampus [17], coupled with hyperconnectivity of the posterior DMN with brain regions outside the network [69]. In the present study, higher intranetwork cohesiveness and lower cross-network connectivity of the DMN were associated with symptomatic improvement (Fig. 3). These findings complement previous observations and indicate that the functional properties of the DMN may play a pivotal role both in disease expression and in treatment response in schizophrenia.

In higher-order networks typical development is associated with decreased intra-network cohesiveness [58] indicating that in the adult brain the nodes of these networks show inherently divergent activity profiles, that likely reflect their engagement in numerous and diverse cognitive operations. The present results implicate mainly the CEN which is known to support multiple aspects of cognitive and attentional control and in goal-directed selection of stimuli and responses [63]. Increased cohesiveness of the CEN has been associated with positive psychotic symptoms in schizophrenia [70] and lower baseline cohesiveness within the subcortical subnetwork of the CEN has been shown to predict antipsychotic response [34]. In accordance with these reports, we found that higher cohesiveness with the frontoparietal and subcortical subnetworks of the CEN was a negative predictor of response.

Based on the multivariate analyses employed here, the predictive value of sMRI features for clinical response was not significant. Nevertheless, the univariate contrasts between responders and non-responders (described in the SI), although not statistical significant in the context of the
multiple comparisons undertaken, implicated the hippocampus bilaterally, and thus align with previous literature [29, 30].

The present study has a number of strengths. To our knowledge, this is the first study to capture the association between neuroimaging measures and clinical improvement across the entire spectrum of psychopathology without being limited to positive symptoms. We used continuous measures of clinical improvement that represent the actual structure of the symptom scores (Supplementary Figure 2). We used a data-driven approach that is conducive to discovery and is more stringent that hypothesis-led analyses that are limited to specific brain regions or networks. We demonstrated the reliability of our main results using alternate brain functional parcellations.

In considering the study limitations, we note that currently available neuroimaging techniques include other modalities (such as diffusion weighted imaging) and other analytic methods (such as graph theory) that were not considered here. We chose to focus on sMRI and rs-fMRI as these represent the most commonly used neuroimaging approaches in psychosis and are also easier to implement in clinical settings because of their relative brevity and lack of requirement for active patient engagement. Sex differences have been reported in incidence, clinical characteristics and mental health service use in schizophrenia [71, 72]. Our sample included a higher proportion of men which is aligned with findings regarding sex differences in service use in the USA [72]. The pattern of findings regarding the association between symptomatic change and resting-state functional connectivity appeared independent of sex, but studies with greater representation of women would be important for confirming our observations. We did not examine drug naïve patients but we ensured that antipsychotic exposure prior to the baseline scan was minimal. Although the daily antipsychotic dose did not correlate with any neuroimaging measure at follow-up, the effect of medication at baseline cannot be conclusively excluded. The results reported here are dependent on the severity of patients’ symptoms which varies across samples and within the same patient over time. Confirmation of the reproducibility of our findings requires replication in larger samples and in longitudinal studies.

Our study provides novel evidence of neuroimaging predictors of the short-term clinical outcome in schizophrenia. These neural network phenotypes can be further interrogated in future studies with regards to their relevance to genetic and molecular mechanisms. For example, the patterns of dysconnectivity observed point to testable hypotheses involving (among several possible alternatives) abnormalities in NMDA receptors and the GABA system [73, 74]. The present findings also have the potential to inform cognitive neuroscientists and clinicians aiming to develop or refine neuromodulatory or psychological interventions that target neural networks most likely to be involved in clinical improvement [75].

Supplementary information is available at MP’s website.

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Compliance with ethical standards

Conflict of interest The authors declare that they have no conflict of interest.

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