Neural substrates of psychosis revealed by altered dependencies between brain activity and white-matter architecture in individuals with 22q11 deletion syndrome

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ABSTRACT

Background: Dysconnectivity has been consistently proposed as a major key mechanism in psychosis. Indeed, disruptions in large-scale structural and functional brain networks have been associated with psychotic symptoms. However, brain activity is largely constrained by underlying white matter pathways and the study of function-structure dependency, compared to conventional unimodal analysis, allows a biologically relevant assessment of neural mechanisms. The 22q11.2 deletion syndrome (22q11DS) constitutes a remarkable opportunity to study the pathophysiological processes of psychosis.

Methods: 58 healthy controls and 57 deletion carriers, aged from 16 to 32 years old, underwent resting-state functional and diffusion-weighted magnetic resonance imaging. Deletion carriers were additionally fully assessed for psychotic symptoms. Firstly, we used a graph signal processing method to combine brain activity and structural connectivity measures to obtain regional structural decoupling indexes (SDIs). We use SDI to assess the differences of functional structural dependency (FSD) across the groups. Subsequently we investigated how alterations in FSDs are associated with the severity of positive psychotic symptoms in participants with 22q11DS.

Results: In line with previous findings, participants in both groups showed a spatial gradient of FSD ranging from sensory-motor regions (stronger FSD) to regions involved in higher-order function (weaker FSD). Compared to controls, in participants with 22q11DS, and further in deletion carriers with more severe positive psychotic symptoms, the functional activity was more strongly dependent on the structure in parahippocampal gyrus and subcortical dopaminergic regions, while it was less dependent within the cingulate cortex. This analysis revealed group differences not otherwise detected when assessing the structural and functional nodal measures separately.

Conclusions: Our findings point toward a disrupted modulation of functional activity on the underlying structure, which was further associated to psychopathology for candidate critical regions in 22q11DS. This study provides the first evidence for the clinical relevance of function-structure dependency and its contribution to the emergence of psychosis.

Abbreviations: FSD, Functional structural Dependency; SDI, Sturctural Decoupling Index; ACC, Anterior Cingulate Cortex; PFC, Prefrontal Cortex; EC, Entorhinal Cortex; AC, Auditory Cortex; 22q11DS, 22q11 Deletion Syndrome.

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1. Introduction

Psychosis is a severe mental illness characterized by hallucinations, delusions and disorganized thoughts. This disorder has a significant impact on the quality of life of patients and their families worldwide (McClellan, 2018; Gore et al., 2011). Extensive research over the years pointed to both genetic and environmental factors contributing to the development of psychosis, but the causes and the underlying neural mechanisms are still largely unknown (Kendler, 2013). The 22q11 deletion syndrome (22q11DS) is a neurogenetic disorder that is among the strongest genetic risk factors for developing psychosis (Biswas and Furniss, 2016). Specifically, approximately 30% to 40% of deletion carriers are diagnosed with schizophrenia by adulthood (Murphy et al., 1999; Lewandowski et al., 2007; Schneider et al., 2014; Schneider et al., 2014). Because 22q11DS is usually diagnosed at a young age due to frequently associated heart or cleft palate malformations (McDonald-McGinn et al., 2015), studies on this population constitute a unique opportunity to map the early stages and the progression of psychosis (Insel, 2010; Lewis and Levitt, 2002). In particular, research on 22q11DS allows to investigate alterations in the neural circuitry that eventually affect sensory and cognitive functions. Indeed, to identify early biomarkers as well as to develop new effective therapies, it is crucial to understand the underlying neurobiological changes associated to the psychopathology.

The brain is constituted of functionally specialized areas that interact together to give rise to perception, cognition and action (Johnson, 2001). Dysconnection of these integrative neural circuits has been consistently proposed as a key mechanism in psychosis (Fornito et al., 2012; Pettersson-Yeo et al., 2011). Accordingly, psychosis is better explained by abnormal interactions between distinct brain areas rather than region specific abnormalities (Stephan et al., 2009). These altered connections have been found at both structural and functional level (McGuire and Frith, 1996; Lawrie et al., 2002; Karbasforoushan and Woodward, 2013; Narr and Leaver, 2015), using diffusion-weighted and functional magnetic resonance imaging (DWI and fMRI), which investigates white-matter pathways and neural activation patterns, respectively. However, by considering the impact of structural and functional connectivity separately, only partial information is yielded which prohibits revealing the complex dynamics of their interaction. Indeed, brain activity is likely to be strongly expressed and constrained by structural white matter pathways (Stiso and Bassett, 2018; Hermundstad et al., 2013; Deco et al., 2013). Therefore, the study of function-structure dependency allows a biologically relevant assessment of behavioral neural mechanisms and has the potential to reveal the neurobiological changes underlying the emergence of psychosis.

The first attempts to characterize the function-structure relationship in healthy subjects involved simple measures of correlation. Specifically, activation time courses from areas with direct structural connections are expected to be statistically more dependent. Indeed, structural connectivity measures have been shown to correlate with brain function (Honey et al., 2010; Honey et al., 2009), and areas that are important hubs of functional connectivity are also found to be key in the structural networks (Sporns et al., 2005). Still, brain structural and function connections do not follow a perfect correspondence. This has been partly explained by indirect structural connections formed on a polysynaptic structural network, as well as the dynamic nature of functional connections that is not reflected in static functional connectivity analysis (Honey et al., 2010; Honey et al., 2009). In individuals with psychosis, the function-structure relationship is likely to be particularly complex. In patients with schizophrenia, findings on function-structure correspondence reported both higher and lower correspondence compared to healthy subjects (van den Heuvel et al., 2013; Skudlarski et al., 2010; Cocchi et al., 2014). Moreover, a considerable number of large-scale abnormalities of functional networks were only in some cases traceable to underlying anatomical changes (Sporns et al., 2005; Cocchi et al., 2014; Crossley et al., 2016). Such inconsistent findings may be partially due to the inability of linear methods, such as correlation, to model the complexity of function-structure relationship. Recent advances in network science (Atasoy et al., 2018) and graph signal processing (Huang et al., 2018) have led to new measures that link regional brain activity and underlying white matter topology, and that were related to different behavioral domains in healthy individuals (Preti and Van De Ville, 2019; Medaglia et al., 2018). Therefore, these measures have a promising potential to provide insights into brain dysfunction and a better understanding of brain communication mechanisms underlying different behavioral domains.

In this study, we applied a methodology that combines the brain function and structure measures (Preti and Van De Ville, 2019) and investigates how much the brain activity in individual regions is exploiting the underlying available white matter structure. In this way, we quantify the regional function-structure dependency (FSD) in patients with 22q11DS and healthy controls using resting-state functional and diffusion MR images. By employing this approach, we detect regions in patients with 22q11DS, for which the changes of functional activity patterns are not supported by accompanying alterations in underlying structural wiring. Consequently, these regions fail to maintain the level of dependency we would observe in healthy controls. Moreover, we further explored how deviation from the expected normal regional FSD could contribute to explain the severity of positive psychotic symptoms. We hypothesized to observe diffuse whole-brain alterations regarding both higher and lower exploitation of underlying structure by the brain activity. Given that previous studies in psychosis and 22q11DS highlighted dysconnectivity in the prefrontal and temporal cortices (Mattia et al., 2018; Ottet et al., 2014), we expected to find individual differences of FSD associated with positive psychotic symptoms mainly in frontal and temporal lobes.

2. Methods and materials

2.1. Participants

Participants were acquired within the on-going cohort of 22q11DS in Geneva which is extensively described in previously published studies (Delavari et al., 2021; Mancini et al., 2020). In this study, we included a total of N = 57 participants with genetically confirmed diagnosis of 22q11DS (31 females, age-span: 16–32) and Nc = 58 healthy control subjects (31 females, age-span: 16–29). Diagnosis of 22q11.2 deletion was confirmed following a quantitative fluorescent polymerase chain reaction performed in the Department of Medical Genetics in Geneva. Subjects in the two groups were carefully matched for age and gender. 22q11DS carriers were recruited as part of a 22q11DS longitudinal study, while healthy controls amongst siblings of the patients or through the Geneva state school system. To assess the presence of psychotic symptoms, the Structured Interview for Prodromal Syndrome (SIPS) has been administered to the patients with 22q11DS by trained clinicians. For this study, we looked at the five positive symptoms subscales (delusional ideas, suspiciousness, grandiose ideas, perceptual abnormalities/hallucinations and disorganized communication). The severity of each symptom was evaluated assigning one score that ranges from 0 (absent) to 6 (severe). For two subjects this information was not available, therefore they were excluded from the severity of psychosis. Participants’ clinical characteristics are further listed in Table 1. Written informed consent was obtained from participants or their parents. The study was approved by the cantonal ethics committee and conducted according to the Declaration of Helsinki.

2.2. MRI acquisition

MRI scans were acquired using a Siemens Trio (n = 108) and a Siemens Prisma-fit (MAGNETOM Trio Upgrade) (n = 7) 3 Tesla scanner. Resting-state fMRI scans were recorded during an 8 min session in which the participants were asked to fixate on a white cross on the screen and
Table 1

Demographic information, presence of psychiatric disorders and drug usage at the moment of the visit in healthy controls and participants with 22q11DS. For the IQ measurements we used Wechsler Intelligence Scale for Children–III (Watkins, 2006) for participants younger than 18 and the Wechsler Adult Intelligence Scale–III (Wechsler, 1955) for the others. To assess the presence of psychiatric disorders we used clinical interview with the patients using the Diagnostic Interview for Children and Adolescents Revised (Reich, 2000), the psychosis supplement from the Kiddie-Schedule for Affective Disorders and Schizophrenia Present and Lifetime version (Kaufman et al., 1997) and the Structured Clinical Interview for DSM-IV Axis I Disorder (First et al., 1997). SD, standard deviation; NA, not available.

| Demographic variables | 22q11DS | Healthy controls | p-value |
|-----------------------|---------|-----------------|---------|
| Number of subjects (F/M) | 57 (31/26) | 58 (31/27) | 0.93 |
| Scanner type: Prima-fit/Trio | 3/54 | 4/54 | 0.71 |
| Average age (SD) | 21.38 (3.7) | 20.93 (4.2) | 0.54 |
| Average IQ (SD) | 72.86 (12.93) | 112.16 (12.81) | <0.001 |
| Average frame-wise displacement after scrubbing (SD) | 0.17 (0.06) | 0.12 (0.04) | <0.001 |

| Anxiety disorder (%) | 31 (54.4%) | 0 | NA |
| Neurodevelopmental deficit hyperactivity disorder (%) | 8 (14%) | 0 | NA |
| Mood disorder (%) | 10 (17.5%) | 0 | NA |
| Schizophrenia spectrum disorders (%) | 5 (8.8%) | 0 | NA |
| More than one psychiatric comorbidity | 13 (22.8%) | 0 | NA |
| Anticonvulsants (%) | 1 (1.7%) | 0 | NA |
| Anidepressants (%) | 2 (3.5%) | 0 | NA |
| Neuroleptic (%) | 8 (14%) | 0 | NA |
| Psychostimulant (%) | 19 | 0 | NA |
| Anxiolytic (%) | 3 (5.2%) | 0 | NA |

not to fall asleep. fMRI images were acquired using a T2-weighted sequence (200 frames) and the following parameters: acquisition matrix = 94 x 128, field of view = 66 x 128, voxel size = 1.84 x 1.84 x 3.2 mm³, 38 axial slices, slice thickness = 3.2 mm, TR = 2400 ms, TE = 30 ms, flip angle = 85°, phase encoding A -> P, descending sequential ordering, GRAPPA acceleration mode with factor PE = 2. Diffusion MRI (dMRI) images were acquired along 30 directions using the following sequence parameters: b = 1000 s/mm², volumetric resolution = 2x2x2 mm³, TR = 8300 ms, TE = 84 ms, flip angle = 90° to 180°, acquisition matrix = 128 x 128, field of view = 25.6 cm, 64 axial slices, slice thickness = 2 mm. A T1-weighted sequence with 192 slices provided anatomical images necessary for the processing of functional and dMRI images (sequence parameters: volumetric resolution = 0.86 x 0.86 x 1.1 mm³, TR = 2500 ms, TE = 3 ms, flip angle = 8°, acquisition matrix = 256 x 256, slice thickness = 1.1 mm).

2.3. fMRI processing

Statistical Parametric Mapping (SPM12, Wellcome Trust Centre for Neuroimaging, London, UK: http://www.fil.ion.ucl.ac.uk/spm/), and functions of the Data Processing Assistant for Resting-State fMRI (DPARSF) were used in order to perform the fMRI preprocessing steps. For each participant, functional images were first realigned over time and spatially smoothed with an isotropic Gaussian kernel of 5 mm full width half maximum (FWHM). Subsequently, anatomical images were coregistered to the functional space and segmented with the SPM12 Segmentation algorithm (Ashburner and Friston, 2005). Brainnetome’s parcellation (https://atlas.brainnetome.org) was resliced to fMRI resolution in order to parcellate the functional images into N = 245 regions of interest (ROD) including cortical and subcortical areas. Nuisance variables were regressed out (6 head motion parameters + other 6, average cerebrospinal fluid, and white matter signal). The first five functional images were excluded and the voxel fMRI time courses were filtered with a bandwidth of 0.01 Hz to 0.1 Hz. In order to obtain regional fMRI time courses, BOLD signals were averaged across all the voxels included in each Brainnetome region. Motion scrubbing (Power et al., 2012) was finally applied for the correction of motion artifacts based on the framewise displacement (FD), which is defined as the sum of the absolute values of the six realignment parameters. When FD was > 0.5 mm, the time point itself, the previous and two consecutive time points were excluded from the analysis. Finally, regional preprocessed time-courses were z-scored. One very small subnucleus of the left thalamus was removed from the current analysis due to its reduced size resulting in no overlap with individual cortices after registration to the individual space and masking (hence, Nk = 245 for the following). Finally, for each participant we obtained a functional connectome constituted out of pairwise Pearson’s correlation between regional time courses. Functional node strength for each region was computed as the sum of absolute correlation values.

2.4. dMRI processing

After visual inspection for motion artefacts, dMRI images were preprocessed and registered to T1 images using Connectome Mapper 3 (https://connectome-mapper-3.readthedocs.io/en/latest/index.html) which is a open-source image processing pipeline software using a combination of libraries such as MRtrix3 (https://www.mrtrix.org/) and FSL (Jenkinson et al., 2012). Subsequently, the following steps of the processing including the warping of the Brainnetome atlas and the generation of the connectome were performed using in-house pipeline. In this section, we report the details of this pipeline.

Firstly, dMRI images were denoised using the MP-PCA algorithm in MRtrix (Veraart et al., 2016; Veraart et al., 2016; Cordero-Grande et al., 2019). Images were corrected for eddy currents and motion using eddy algorithm in FSL (Andersson and Sotiropoulos, 2016). Then, the T1 anatomical images were registered to diffusion space using a non-linear registration method (ANTS) with the algorithm SyN (Avants et al., 2008) and were segmented into cortical gray matter, subcortical gray matter, white matter and CSF using FreeSurfer. The Brainnetome atlas was warped from MNI to anatomical subject-space and down-sampled to dMRI resolution using FSL-FNIRT (Andersson et al., 2007). Fiber orientation distribution was estimated using a constrained spherical deconvolution with single-shell single-tissue response function using the sdc algorithm in MRtrix (Tournier et al., 2007). Probabilistic fiber tracking was applied to reconstruct 106 streamlines using the tcggk function in MRtrix with the following parameters: step size of 0.5, angle of 45, maximum length of 250 mm, a cutoff of 0.06. Anatomically-Constrained Tractography framework during tracking was also used (Smith et al., 2012). Subsequently, the tractogram was filtered using spherical-deconvolution informed filtering of tractograms 2 (SIFT2) (Smith et al., 2015) for each subject. The structural connectomes were obtained by calculating the number of streamlines linking every pair of the Nk = 245 regions defined for the functional analysis, divided by the total number of streamlines. Finally, in order to quantify the structural node strengths, for each participant we computed the sum of each row of the structural connectome.

2.5. Function-structure dependency

For each participant, we calculated the structural decoupling index (SDI) for every region, using the graph-signal-processing methodological pipeline detailed in (Freti and Van De Ville, 2019). In brief, structural harmonics; (i.e., brain patterns that most naturally encode the wiring architecture), were obtained by the eigendecomposition of the normalized Laplacian of an individual’s structural connectome. This yielded Nk Laplacian eigenvectors, so called harmonic components, each of which is associated to an eigenvalue that can be interpreted as a graph frequency value. Harmonic components with low graph frequencies are, by construction, ‘easier’ to express on the structural connectome; i.e., they represent global brain patterns along the main
geometrical axes such as anterior-posterior or left–right. While the components associated with higher graph frequencies are capturing more complex and localized patterns. We subsequently projected the individual functional data, for each time point, onto the individual structural harmonics and applied a graph signal filtering that decomposes the activity signal into two parts: one expressed on low-frequency structural harmonics (and therefore more aligned with structure), and the other one on the complementary high-frequency ones (hence, more detached from structure). The cut-off frequency of the ideal low-/high-pass filter is defined by an equal-energy split of the energy spectral density of each subject. The norm across time of aligned and detached signal portions was computed and their ratio yielded the SDI. It is noteworthy the term structural decoupling index is not to be confused with use of the term “coupling” in the field of neuro-modulation. SDI quantifies the absence of function-structure dependency in each region. Therefore, brain areas with an SDI > 1 are regions whose activity signals are relatively more divergent from underlying structural pathways and thus have lower FSD, and the opposite occurs for regions with SDI < 1.

2.6. Statistical analysis

In order to identify for which brain regions of the brain nodal functional connectivity strength, nodal structural connectivity strength, or the SDI significantly differ in participants with 22q11DS and healthy controls, we used a two-sample unpaired t-test for normal distributions, or the non-parametric test named Wilcoxon rank sum test in case of non-normal distributions. (More information regarding the test for normality is presented in supplementary materials. Table S3) We corrected the results for multiple comparisons using False Discovery Rate (FDR) correction (Benjamini and Hochberg, 1995). Age and motion (average FD after scrubbing) were included as nuisance regressor in group comparisons. Further, to investigate the implication of individual differences of SDI in the severity of positive psychotic symptoms, we applied a behavioral Partial Least Squares Correlation (PLS-C) analysis (McIntosh and Lobaugh, 2004; Krishnan et al., 2011) using myPLS toolbox (https://github.com/danizoeller/myPLS). PLS-C is a multivariate approach that aims at finding linear combinations of the original data (i.e., latent components) maximizing the covariance between brain and behavioral data, represented in our case by the SDI in patients with 22q11DS and seven behavioral variables corresponding to five positive psychotic symptoms subscales scores, sex, and age. Before entering the SDI and the behavioral measures into PLS-C, motion (average FD after scrubbing) was regressed out for each variable separately. As a result, we obtained brain and behavioral saliences implementing a bootstrap procedure consisting of 500 random samples with replacement. The brain pattern is visualized choosing bootstrap ratio scores (BRS) that are in absolute value >2.3, corresponding to a confidence level of approximately 95%. Instead, the behavioral saliences are presented as bar plots with the related 95% confidence interval bars according to each bootstrap distribution. In order to assess the effectiveness of SDI, a PLS-C has also been performed for both the functional and structural connectivity node strength measures separately.

3. Results

3.1. Function-structure dependency

The multimodal analysis in healthy controls (HCs) and patients

![Comparison between the brain SDI gradient of healthy controls and participants with 22q11DS.](image)

**Fig. 1.** Comparison between the brain SDI gradient of healthy controls and participants with 22q11DS. (A) The binary logarithm of the average SDI calculated for the healthy controls. (B) The binary logarithm of the average SDI calculated for the participants with 22q11DS. In A) and B) primary sensory cortices present the strongest FSD (low values of SDI - dark blue points), frontal cortex exhibits moderate FSD (green and light blue points), while inferior temporal cortex and subcortical regions present the weakest FSD (high values of SDI - red points). Thus, the gradient of FSD as in sensory cortices > frontal cortices > subcortical cortices, is consistently present in both groups. (C) Brain map of the differences between the binary logarithm of average SDI of healthy controls and the binary logarithm of the average SDI of participants with 22q11DS. Compared to healthy controls, participants with 22q11DS present stronger FSD in occipital, inferior temporal, superior frontal lobes and subcortical areas (yellow regions). Concurrently, FSD is weaker mainly in prefrontal and superior temporal cortices (pink regions). (D) Statistically significant SDI group differences after correction for multiple comparisons. In C) and D), the dimension of the points depicts the magnitude of the regional SDI difference in binary logarithm form, which could be viewed as the relative difference across the two groups (detailed in supplementary material Table S.3). (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)
generated a macroscale gradient of SDI over the whole brain (Fig. 1A and 1B). The cortical gradient revealed regions in visual, sensory, motor and auditory cortices had the lowest SDI values (dark blue nodes) indicating a strong alignment between function and structure (high FSD). On the contrary, cortices dedicated to higher-level cognitive function mainly located in the frontal and temporal lobes (light blue, yellow and light red nodes) presented relatively higher SDI values, indicating relatively lower FSD. These results were in line with findings in previous work (Preti and Van De Ville, 2019), except that the inclusion of subcortical brain regions showed that the FSD further decreases for regions such as amygdala, hippocampus, thalamus and basal ganglia (highest SDI values - dark red nodes), for which the functional activity is the least constrained by the underlying anatomical backbone.

Table 2
Brainnetome atlas regions for which the average SDI is significantly different between healthy subjects and participants with 22q11DS. The first column indicates the region’s location in the brain, while the second column indicates the brain region’s name (BA = Brodmann Area) with the correspondent region’s IDs of the Brainnetome atlas (third column). The average SDI value for healthy controls (SDI_HC), the average SDI value for participants with 22q11DS (SDI_22q11DS), the adjusted p-values (q-value) after correction for multiple comparisons are also listed in the following columns. Finally, we report in the last columns the average structural nodal strength which consists of sum of all connections for a certain region. Structural nodal strength was reported for healthy controls (SC_HC) and for 22q11DS (SC_22q11DS) with the respective adjusted p-value (q-value) after multiple comparisons. For region with significantly different SDI values, the average correlation coefficient between SDI and motion (in terms of mean frame-wise displacement after scrubbing) is reported.

| Region’s location          | Region’s name (hemisphere)         | Region’s ID | Region’s ID | Region’s ID | Region’s ID | Region’s ID | Region’s ID |
|----------------------------|-------------------------------------|-------------|-------------|-------------|-------------|-------------|-------------|
| Superior Frontal Gyrus     | medial BA 8 (L)                     | 2           | 0.92        | 0.80        | 0.0153      | 0.0048      | 0.0052      | >0.05       |
| Orbital Gyrus              | lateral BA 11 (L)                   | 45          | 0.92        | 0.82        | 0.0352      | 0.0045      | 0.0049      | 0.0428      |
| Paracentral Lobule         | BA 4 (R)                            | 68          | 0.67        | 0.58        | 0.0195      | 0.0042      | 0.0043      | >0.05       |
| Inferior Temporal Gyrus    | intermediate lateral BA 20 (L)      | 95          | 1.43        | 1.14        | 0.0133      | 0.0015      | 0.0021      | 0.0000      |
| Superior Parietal Lobule   | postcentral BA 7 (R)                | 132         | 0.93        | 0.78        | 0.0071      | 0.0038      | 0.0041      | >0.05       |
| Inferior Parietal Lobule   | caudal BA 39 (L)                    | 135         | 0.83        | 0.73        | 0.0150      | 0.0053      | 0.0057      | >0.05       |
| Insular Gyrus              | medial BA 5 (L)                     | 150         | 0.82        | 0.72        | 0.0325      | 0.0040      | 0.0037      | >0.05       |
| Lateral Occipital Cortex   | middle occipital gyrus (L)         | 199         | 0.69        | 0.59        | 0.0116      | 0.0056      | 0.0053      | >0.05       |
| Hippocampus                | inferior occipital gyrus (L)       | 200         | 0.68        | 0.58        | 0.0195      | 0.0049      | 0.0048      | >0.05       |
| Basal Ganglia              | ventral caudate (R)                | 216         | 0.78        | 0.68        | 0.0228      | 0.0052      | 0.0051      | >0.05       |
| Thalamus                   | medial pre-frontal thalamus (R)    | 232         | 2.62        | 2.36        | 0.0149      | 0.0013      | 0.0015      | 0.0006      |
| Lower average FSD in 22q11DS (SDI_22q11DS < SDI_HC) | | | | | | | |
| Region’s location          | Region’s name (hemisphere)         | Region’s ID | Region’s ID | Region’s ID | Region’s ID | Region’s ID | Region’s ID |
| Middle Frontal Gyrus       | medial BA 45 (L)                    | 35          | 0.84        | 1.03        | 0.0022      | 0.0025      | 0.0026      | >0.05       |
| Inferior Frontal Gyrus     | medial BA 17 (R)                    | 36          | 0.85        | 1.02        | 0.0071      | 0.0025      | 0.0025      | >0.05       |
| Orbit Gyrus                | lateral BA 12/17 (R)                | 51          | 0.57        | 0.71        | 0.0003      | 0.0048      | 0.0054      | 0.0000      |
| Precentral Gyrus           | BA 6 (L)                            | 61          | 0.71        | 0.79        | 0.0035      | 0.0037      | >0.05       |
| Auditory Cortex            | TE1.0 and TE1.2 (L)                | 73          | 0.56        | 0.66        | 0.0078      | 0.0039      | 0.0039      | >0.05       |
| Superior Temporal Gyrus    | lateral BA 38 (R)                   | 78          | 0.89        | 1.03        | 0.0230      | 0.0027      | 0.0028      | >0.05       |
| Inferior Temporal Gyrus    | extreme lateralventral BA37 (R)     | 91          | 0.75        | 0.86        | 0.0484      | 0.0030      | 0.0030      | >0.05       |
| Insular Gyrus              | dorsal agranular insula (L)        | 153         | 0.66        | 0.75        | 0.0149      | 0.0054      | 0.0051      | >0.05       |
| Cingulate Gyrus            | rostralventral BA 24 (L)            | 178         | 0.69        | 0.83        | 0.0027      | 0.0066      | 0.0065      | >0.05       |
| Medial Frontal Gyrus       | medial BA 7 (L)                     | 147         | 0.77        | 0.87        | 0.0149      | 0.0034      | 0.0031      | >0.05       |
| Basal Ganglia              | ventral thalamus (R)               | 151         | 0.46        | 0.54        | 0.0025      | 0.0069      | 0.0059      | 0.0000      |
| Precentral Gyrus           | BA 3 (L)                            | 153         | 0.66        | 0.75        | 0.0149      | 0.0054      | 0.0051      | >0.05       |

| Framewise displacement correlation with SDI |
|---------------------------------------------|
| Correlation coefficient (r-value), Average (standard deviation) |
| Areas with significantly different SDI in the groups |
| HC | 22q11DS | P-value |
| 2.46e-17 (1.02e-16) | 4.32e-18 (7.63e-17) | 0.3502 |

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3.2. Comparing participants with 22q11DS against healthy controls

The map of whole brain average SDI differences (defined as SDI\textsubscript{HC} - SDI\textsubscript{22q11DS}) is presented in Fig. 1C. Comparing the average regional SDI between patients with 22q11DS and HCs, 48 out of 245 brain regions remained significantly different across the groups after multiple comparisons correction (Fig. 1D). Regions with significantly higher FSD in patients with 22q11DS (yellow nodes) were mainly clustered in inferior temporal, superior parietal, lateral occipital cortex and subcortical areas. Conversely, prefrontal regions along with cingulate cortex presented a weaker FSD (pink nodes) in participants affected by the deletion when compared with HCs. Table 2, presents the average SDI values in both groups and the respective p-values after multiple comparison. Additionally, for the depicted regions, correlation of motion variable (frame-wise displacement after scrubbing) and SDI within each region in both groups of 22q11DS and healthy controls was calculated. Average correlation coefficient did not differ between patients and healthy controls (Wilcoxon rank sum test, p-value = 0.3502). A more detailed investigation of contribution of motion to SDI is provided in supplementary materials.

3.3. Association with positive psychotic symptoms

PLS-C behavioral analysis resulted in one significant latent component (LC1, p = 0.017), which captured a strong positive effect of all the five positive psychotic symptoms, a small negative effect of gender and no effect of age (Fig. 2B). Thus, the corresponding brain saliences (Fig. 2A) mainly display a pattern in which SDI is broadly associated with the severity of psychosis. The brain activity of patients experiencing higher scores of psychotic symptoms is less dependent on the white matter structure in the anterior cingulate cortex (ACC), thalamus and parietal lobule (red regions in Fig. 2A). On the contrary, brain activity is more aligned with structure in inferior and superior temporal gyrus, parahippocampal gyrus, and putamen (blue regions in Fig. 2A).

![Fig. 2.](image) Individual differences in SDI are significantly associated with the severity of positive psychotic symptoms in participants with 22q11DS. Brain and behavioral saliences relative to the significant LC resulted from the PLS-C analysis are displayed. (A) The pattern of brain saliences shows that compared to deletion carriers experiencing mild or no positive psychotic symptoms, the ones presenting severe symptoms exhibit stronger FSD in the parahippocampal and inferior temporal gyrus (blue regions). Concurrently, FSD is weaker mainly in the cingulate gyrus and in the parietal cortex (red regions). (B) The design saliences reveal a strong positive effect of all the five positive symptoms subscales, a small effect for gender and no effect for age. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)
Each specific atlas region with the correspondent salience value is listed in Table 3. Additionally, we conducted a supplementary PLS including a variable for consumption of psychostimulants, which did not produce a robust loading (Fig. 4S).

3.4. Contribution of unimodal measures

In terms of node strengths of the functional connectome, no significant group difference was found. On the contrary, differences among the node strengths of the structural connectome were found across all the brain (75 regions out of 245). In Table 2, the structural nodal values and the respective p-values after multiple comparisons correction are listed for each region in which the SDI values were significantly different. The regions in which we found different structural nodal strengths between the two groups are visible in Fig. 1S of supplementary materials. PLS-C analysis performed for functional and structural node strength measures separately in both cases resulted in no significant components. Of note, to explore an approach already used in previous literature we further examined the connectomes in terms of connectivity measures. Thus, we performed an independent t-test analysis for each ROI-to-ROI connection within the functional and structural connectomes. Results for group differences among structural and functional connections, after correcting for multiple comparisons, are reported in Fig. 2S and 3S.

4. Discussion

In line with previous findings in healthy adults reported in Preti et al. (Preti and Van De Ville, 2019); our analysis revealed a major spatial gradient ranging from sensory-motor regions (stronger FSD) to regions involved in higher-order function (weaker FSD); see Fig. 1A and 1B. Similarly, multiple studies have already reported a comparable gradient that spans from strongly aligned ‘unimodal’ sensory cortices to weakly aligned ‘trans-modal’ cortices using different measures (Margulies et al., 2016; Vázquez-Rodríguez et al., 2019; Baum et al., 2020; Paquola et al., 2019). This hierarchical organization of the brain has been proposed to support increasing levels of flexible and dynamic processes. Accordingly, polysynaptic indirect connections are more likely to be involved in higher-order integrative processes (Buckner and Krienen, 2013).

Brain areas at the apex of the hierarchy may activate more in synchrony, not only as a consequence of direct signaling between them, but also by being driven by common inputs received from the rest of the brain. Consequently, the simultaneous functional activation of regions that are not structurally linked leads to a weak FSD (Bettinardi et al., 2017; Damoiseaux and Greicius, 2009). On the contrary, in sensory regions the functional activity could be more directly supported by the underlying white matter pathways because of the need for fast reactions to internal or external stimuli (Mesulam, 1998). Furthermore, by adding the subcortical regions we observed that functional activity is even less constrained by anatomical backbone compared to frontal and temporal trans-modal cortices. These regions are involved in complex and polysynaptic circuits (Utter and Basso, 2008; Berridge and Kringlebot, 2015; Janak and Tye, 2015) and the weakest alignment observed further supports the hypothesis that a decrease in function-structure alignment consents increasing flexible and dynamic processes (Mesulam, 1998).

Current knowledge converges on the idea of a hierarchical organization of functional-structural dependency. However, the divergence from a normal FSD within one region and its pathological implications are not well understood yet (Suárez et al., 2020). In the present study, our analysis revealed a similar FSD gradient throughout the brain for both groups of 22q11DS and HCs, which confirmed that the dominant FSD pattern is reflecting a robust organizational principle of the brain. However, when comparing regional FSD between 22q11DS and HCs by means of structural decoupling index, we found a significant difference in several brain areas. A considerable number of regions with lower FSD in 22q11DS were located in the prefrontal cortex (PFC) and ACC, which are well known to play a crucial role in executive function, attention, memory and emotional regulation (Miller and Cohen, 2001). In fact, better executive function has been associated with higher prefrontal FSD (Medaglia et al., 2018). In this context, pruning occurring during normal development of the PFC, may lead to more efficient white matter mediation to support the complex functionality of PFC (Kolb and Gibb, 2011; Hooper et al., 2004). Therefore, sufficiently high FSD in prefrontal areas could be indicative of a well-matured cortex (Mills et al., 2014; Schlegel et al., 2012; 2012). Consequently, the abnormally low FSD detected in 22q11DS in this study, could point to the structural impairments resulting from an abnormal pruning process and immature PFC that does not form the optimal structural connections to support complex executive functions. Structural impairments in prefrontal cortex and abnormal pruning (Shashi et al., 2012; Schaer et al., 2009) have been already documented in 22q11DS (Radoeva et al., 2012) and have been further associated with behavioral dysfunctions and psychotic symptoms (Schreiner et al., 2014). Additionally, our analysis detected abnormally high FSD in striatal area of patients with 22q11DS, which was associated with the severity of psychotic symptoms. Indeed, it has

Table 3

Brainnetome atlas regions for which the SDI values were significantly associated to the correspondent region is also listed (fourth column): positive values indicate that deletion carriers with more severe psychotic symptoms present weaker FSD, while negative values indicate they present stronger FSD.

| Higher FSD in deletion carriers experiencing positive psychotic symptoms | Region’s Location | Region’s name (hemisphere) | Region’s ID | Brain salience |
|---|---|---|---|---|
| Superior Temporal | Gyrus | BA 41/42 (R) | 72 | 2.48 |
| | | TE1.0 and TE1.2 (R) | 74 | 2.34 |
| | | rostral BA 22 (R) | 80 | 2.36 |
| Inferior Temporal | Gyrus | rostral BA 20 (L) | 93 | 3.32 |
| | | intermediate lateral BA 20 (L) | 95 | 2.47 |
| | | intermediate lateral BA 20 (R) | 96 | 3.09 |
| Parahippocampal | Gyrus | rostral BA 35/36 (R) | 110 | 3.54 |
| | | BA 28/34 (entorhinal cortex) (L) | 115 | 3.53 |
| Lower FSD in deletion carriers experiencing positive psychotic symptoms | Region’s Location | Region’s name (hemisphere) | Region’s ID | Brain salience |
|---|---|---|---|---|
| Superior Frontal | Gyrus | BA 6 (R) | 19 | 2.32 |
| Paracentral Lobule | | BA1/2/3 (lower limb region) (R) | 66 | 2.67 |
| Parahippocampal | Gyrus | posterior parahippocampal gyrus (L) | 113 | 3.34 |
| Inferior Parietal Lobule | Gyrus | area 7H (medial PPHC) (R) | 120 | 2.63 |
| Precentral | Gyrus | BA 40 (R) | 142 | 2.78 |
| Premotor | Gyrus | rostral BA 24 (L) | 177 | 4.04 |
| Cingulate | Gyrus | rostral BA 24 (R) | 178 | 3.38 |
| | | pregenual BA 32 (L) | 179 | 3.14 |
| | | caudal BA 23 (L) | 185 | 2.89 |
| | | caudal BA 23 (R) | 186 | 4.89 |
| Thalamus | | sensory thalamus (R) | 234 | 3.70 |
been already suggested that the abnormal PFC maturation in 22q11DS is driven by dopaminergic dysfunction (Schaer et al., 2009). Notably, insufficient cortical dopamine combined with sub-cortical hyper-dopaminergic state is one of the most accepted mechanism in psychosis (Lodge and Grace, 2007) and it has been recently indicated in 22q11DS (Delavari et al., 2021). In that regard, abnormally high FSD within the striatal area could be indicative of a maximal subcortical dopaminergic activation that saturates the underlying wiring structures.

In addition, our results pointed towards an association between the presentation of positive psychotic symptoms and the decrease of FSD in ACC within participants with 22q11DS. Indeed, ACC has been consistently associated with the presence and severity of psychotic symptoms in both functional and structural neuroimaging studies conducted in patients with 22q11DS (Rhins et al., 2013; Scarlatti et al., 2016; Sandini et al., 2018; Tomescu et al., 2014; Padula et al.,), as well as idiopathic schizophrenia (Allen et al., 2008; Menon, 2011). Reductions in white matter tracts observed in the ACC of patients with 22q11DS could lead to functional reorganization that finally results in the development of alternative activation patterns, detached from the underlying structural connections. Given that ACC, as part of the salience network, plays an active role in detecting salient stimuli, the detachment we observed in ACC may point towards irrelevant activations, resulting in misattributions of salience (Kapur, 2003). On the other hand, our results further denoted a higher alignment in the entorhinal cortex (EC) of patients with 22q11DS with more severe psychotic symptoms. Notably, EC is part of the novelty detection circuit (Witter et al., 2000), it is connected to ACC via cingulum (Rolls, 2019) and it mediates the input and output of the hippocampus. EC mainly holds sensory information while the hippocampus compares it with internal representations (Falkai et al., 2000; Prasad et al., 2004). Previous studies in schizophrenia found reductions in volume of EC (Arnold, 1999; Bettinardi et al., 2017) and higher activation in parahippocampal gyrus has been previously linked to psychosis (Delavari et al., 2021; Zöller et al., 2017; Boley et al., 2014; Friston et al., 1992). This finding may suggest an over-engagement in the local circuits of hippocampal complex, leading to a disruption of the novelty detection processes. This, along with misattribution of salience has been proposed as a potential mechanism contributing to distortion of reality that is observed in psychosis (Lisman and Otmakhova, 2001).

The relatively elevated FSD values in sensory regions of patients with 22q11DS, may implicate that brain activity is more restricted to local structural circuits and, therefore, these regions are less integrated with higher order cognitive areas. The resulting segregation of the superior parietal and inferior temporal lobe could explain the wide-ranging impairments in sensory domains and visuospatial abilities observed in 22q11DS (Larsen et al., 2019; Bostelmann et al., 2016; Attout et al., 2017). Our analysis isolated the primary auditory cortex (AC), by detecting a lower FSD in 22q11DS as compared to HC, which is an exception among other sensory areas. Several studies in 22q11DS have shown robust functional hyperactivity in AC, which strongly correlates with the emergence of auditory hallucinations (Mancini et al., 2020; Ferri et al., 2018; Li et al., 2017). Intriguingly, stronger AC alignment was detected in patients presenting more severe positive psychotic symptoms, thus the lower FSD in AC of patients with 22q11DS, was driven by less symptomatic patients. This could be pointing towards a trait, where compensatory integration of auditory circuits has a protective role against positive psychotic symptoms.

Overall, our findings showed abnormal function-structure dependencies within and between different networks and individuals with 22q11DS. Notably, when analyzing separately the nodal measures for structural and functional connectomes, we found no significant results. Therefore, the FSD was unique in providing significant associations with psychosis psychopathology. The present study provides first evidence of the clinical relevance of the FSD, highlighting changes that are not otherwise identifiable with a nodal functional or structural analysis alone. These results suggest that the emergence of psychosis may be more tightly related to a disrupted ability of brain networks to modulate functional activation on the underlying white-matter pathways, rather than to separate alterations of nodal structural or functional connections. It is noteworthy that a post-hoc analysis examining the group differences in the ROI-to-ROI connectivity, resulted in statistically significant differences reported in previous literature that explored functional and structural connectivity in participants with 22q11DS. However, here we showed that nodal measures for structural or functional scans, taken separately, fail to reflect the changes expected in participants with 22q11DS. Nevertheless, by combining functional and structural information into the integrative FSD framework led to meaningful results at a nodal level. Therefore, we suggest that the methods assessing functional structural dependency hold a potential to robustly detect differences when performing a regional level analysis.

4.1. Conclusions and further perspectives

By combining information from functional and diffusion MR imaging in a single unified framework, we showed that brain function is differently constrained by the anatomical structure in 22q11DS. However, the findings of this study present some limitations that are worthy to be considered. Firstly, morphometric alterations in 22q11DS impose a limitation on the conducting analysis in a normalized template space. To overcome this, we have calculated the SDI in the native subject space. However, atlas realignment are not ideally formed to address morphological alterations in this population. Furthermore our data consists of dMRI images acquired along 30 directions using a b-value of 1000 s/mm, which are the minimal technical requirements to obtain an optimal tensor estimation (Calamoneri et al., 2018). Certainly, the latest available DSI or HARDI sequences use more gradient directions and higher b-values which results in better reconstruction of crossing fibers (Tournier et al., 2004). Additionally, anatomically-constrained tractography (ACT) and spherical-deconvolution informed filtering of tractograms (SIFT) algorithms were adopted in the tractographic reconstruction of the SC, as shown to produce more biologically realistic connectomes (Smith et al., 2012) and reduce biases in streamline densities (Smith et al., 2015), respectively, leading to more interpretable results. Nonetheless, this correction is not perfect and brain network metrics computed on SC were previously shown to be affected by the methods used to perform tractogram bias correction (Yeh et al., 2016). In particular, a bias towards over-represented long fibers might lead to a decreased coupling for localized brain systems involving short connections, which, however, does not seem to occur in our case and in previous analyses of SDI (e.g., highly coupled visual / auditory systems; Preti and Van De Ville, 2019; Griffa et al., 2021). Nonetheless, the effect of different SC processing choices could be worth evaluating in future work.

Furthermore, tractography reconstructions can be influenced by a number of factors, such as movement, that might lead to confounds. To overcome this, we employed state-of-the-art methodologies for the preprocessing of dMRI data, and we regressed out motion from our analysis. However, average within-scanner was significantly higher in patients with 22q11DS, when compared to healthy controls. We have tried to further control for the effect of motion by regressing-out motion variable from our analysis. We have further conducted a series of post-hoc analysis to ensure the results are not driven by the difference of motion across the two groups (described in detail in supplementary material). Moreover, SDI methodology gives a nodal summary measure for each region. Therefore, this highlights regions with overall changes, without providing information on specific connection alterations. Contrary to the previous work (Preti and Van De Ville, 2019) in which the relationship between brain structure and function was investigated using the structural connectome at the group level, in the present study the structural connectome of each subject has been taken into account. This permitted characterizing the alignment of brain function and anatomy at the individual level and to correlate for each patient the FSD pattern with the severity of positive psychotic symptoms.
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To the best of our knowledge this is the first study to assess the SDI alterations in a pathologic group. We have quantified the differences of functional-structural dependency across the brains of patients with 22q11DS and further evaluated how these changes are correlated with positive psychotic symptoms. Indeed, the method employed here does not provide direct information regarding the underlying neurobiological changes. However, our results revealed higher FSD in patients with 22q11DS, occurring mainly in hyperactive regions known to be segregated from higher order cortices. Conversely, lower FSD in patients with 22q11DS coincided with regions that are known to go through abnormal pruning and form suboptimal connections. Future studies on SDI alterations in pathologic populations may provide further insight into the biological correlates of SDI. Moreover, studies characterizing the age-dependent trajectories of SDI may shed light on how dependency of functional activity over structural underlay changes across the development.

CRediT authorship contribution statement

Karin Bortolin: Conceptualization, Investigation, Methodology, Formal analysis, Software, Writing – original draft, Visualization. Farzad Delavari: Conceptualization, Investigation, Methodology, Formal analysis, Software, Writing – original draft, Visualization. Maria Giulia Preti: Conceptualization, Methodology, Software, Writing – review & editing. Corrado Sandini: Conceptualization, Writing – review & editing. Valentina Mancini: Investigation, Writing – review & editing. Emeline Mullier: Methodology, Software, Writing – review & editing. Dimitri Van De Ville: Conceptualization, Resources, Methodology, Writing – review & editing, Supervision, Project administration, Funding acquisition. Stephan Eliez: Conceptualization, Investigation, Resources, Writing – review & editing, Supervision, Project administration, Funding acquisition.

Declaration of Competing Interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Appendix A. Supplementary data

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