Disruption of superficial white matter in the emotion regulation network in bipolar disorder

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ABSTRACT

Bipolar disorder (BD) is characterized by emotion dysregulation and involves changes in the gray matter (GM) and white matter (WM). Although previous diffusion tensor imaging (DTI) studies reported changes in the diffusion properties of the deep WM (DWM) in BD patients, the diffusion properties of the superficial WM (SWM) are rarely investigated. In this study, we tried to determine whether the diffusion parameters of the SWM were altered in BD patients compared to controls and whether the changes were associated with the disrupted emotion regulation of the BD patients. We collected DTI data from 37 BD patients and 42 gender- and age-matched healthy controls (HC). Using probabilistic tractography, we defined a population-based SWM mask based on all the subjects. After performing the tract-based spatial statistical (TBSS) analyses, we identified the SWM areas in which the BD patients differed from the controls. This study showed significantly reduced fractional anisotropy \((FA_{SWM})\) in the BD patients compared to the HC in the bilateral dorsolateral prefrontal cortex (dlPFC), ventrolateral prefrontal cortex (vlPFC), medial prefrontal cortex (mPFC), and the left parietal cortex. Moreover, compared to the controls, the BD patients showed significantly increased mean diffusivity \((MD_{SWM})\) and radial diffusivity \((RD_{SWM})\) in the SWM in the right frontal cortex. This study presents altered cortico-cortical connections proximal to the regions related to the emotion dysregulation of BD patients, which indicated that the SWM may serve as the brain's structural basis underlying the disrupted emotion regulation of BD patients. The disrupted \(FA_{SWM}\) in the parietal cortex may indicate that the emotion dysregulation in BD patients is related to the cognitive control network.

1. Introduction

Bipolar disorder (BD) (Chang et al., 2013) is characterized by dysfunction in emotion regulation (Phillips and Swartz, 2014), which refers to the ability of an individual to modulate their response to emotional stimuli (Frank et al., 2014). Dysregulation of the emotions may lead to mood disorders (Townsend et al., 2013). Previous studies revealed a neural circuit (Phillips and Swartz, 2014; Phillips et al., 2008), comprising cortical regions that include the ventrolateral prefrontal cortex (vPFC), dorsolateral prefrontal cortex (dPFC), and medial prefrontal cortex (mPFC) as well as subcortical regions, including the amygdala and striatum, which may underlie disrupted emotion.
regulation in BD patients. Recent meta-analyses (Zilverstand et al., n.d.; Buhle et al., 2014) that summarized previous studies of emotion regulation suggested that this emotion regulatory network largely overlaps with the cognitive control network and indicated that abnormalities in the parietal regions, especially in the inferior parietal (IP) and superior parietal (SP) gyrus, may also indicate a deficit in emotion regulation.

Although the pathology of BD remains unknown, white matter (WM) abnormality is believed to be related to the disease mechanism (Mahon et al., 2010). Brain WM can be classified into two systems, the superficial WM (SWM) and the deep WM (DWM), based on differences in their morphological features, including size, arrangement, and distribution (Oishi et al., 2008). WM studies in BD have primarily considered the DWM, which comprises large WM bundles and enables long-range connections between gray matter (GM) regions. In contrast, the SWM enables local connections between cortico-cortical regions. The SWM is located below the cortex (Oishi et al., 2008) and consists of short cortico-cortical fibers, such as U-shaped and intra-lobar fibers (Nazeri et al., 2013).

Although less is known about the specific functions of the SWM, several studies have indicated that the SWM is related to human cognitive functions. For example, Nazeri et al. found that the SWM contributes to cognitive performance in patients with schizophrenia (Nazeri et al., 2013) and in aging populations (Nazeri et al., 2015). Recently, Liu et al. (Liu et al., 2016) reported that in temporal lobe epilepsy the SWM serves as an important link between the functional and structural networks. These findings suggest an important role for the SWM, but an effective tractography for SWM fibers has not yet been fully identified.

Because of the complexity of the brain’s cortical folding patterns (Chen et al., 2013; Nie et al., 2012), compared to DWM, the SWM is more complex, which impede reconstructing long-range connections near the cortical regions (Reveley et al., 2015) using tractography. For example, the fractional anisotropy (FA) and curvature measured at the white/gray matter boundary (WGB) are respectively lower and sharper than those measured in the DWM. Additionally, in a given fiber bundle, it is difficult to distinguish between the fibers (Nazeri et al., 2013) that belong to the DWM and those that belong to the SWM. However, with more advanced fiber-tracking algorithms (Behrens et al., 2007), a researcher can use “waypoint mask” or “exclusion mask” to constrain fibers to those near the cortex and remove the long-range fibers passing through the exclusion regions, if the seeds were selected at the WGB.

In this study, our goals were to determine whether the SWM is altered in BD patients compared to healthy controls (HC) and to determine whether these SWM alterations are associated with the disrupted emotion regulation of BD patients. In the calculations, we adopted probabilistic tractography (Behrens et al., 2007) to reconstruct a population-based SWM mask and used a tract-based spatial statistics (TBSS) (Smith et al., 2006) analysis to assess the differences of diffusion parameters in the SWM regions between the BD patients and the HC group.

2. Methods

2.1. Subjects

Thirty-seven BD subjects were recruited from the First Affiliated Hospital of Jinan University (AHJU), Guangzhou. The eligible subjects for this study were right-handed, aged 18–55 years old, and able to read and write. All the patients (26 M/11 F, aged 18–51 years old) completed the Young Mania Rating Scale (YMRS) (Young et al., 1978) and the Hamilton Depression Scale (HAMD) (Whisman et al., 1989). The YMRS was used to measure the manic degree of the BD patients in the week before the scan and the HAMD was used to detect each individual’s depression state. The inclusion criteria for the BD group were (i) no long-term history of medication usage and (ii) a YMRS score < 12 and HAMD score > 21. The exclusion criteria for the patients were as follows: patients with other comorbid DSM-IV Axis-I psychiatric disorders or a history of neurological disorders, organic brain disorders, cardiovascular diseases, pregnancy, and alcohol or substance abuse. We also recruited forty-two healthy subjects (26 M/16 F, aged 18–52 years old) as the control group in the hospital. The exclusion criteria for the HC were the same as that of the BD patients. In addition, to remove comorbidities we excluded subjects who had a history of or first-degree relatives with psychiatric illness. Table 1 lists the demographic characteristics of the BD patients and the HC group and the clinical performance of the BD patients. This study was approved by the Institutional Review Board of the First Affiliated Hospital of Jinan. Written informed consent was obtained from each subject prior to the study.

2.2. Data acquisition

All MRI data were acquired on a 3 T GE MR750 scanner with 8-channel head coil. The DTI data were obtained using a single-shot diffusion-weighted EPI sequence with the following parameters: repetition time (TR) = 8,000 ms, echo time (TE) = 68 ms, field of view (FOV) = 256 mm × 256 mm, data matrix = 128 × 128, voxel size = 2 × 2 × 2 mm³, slice thickness = 2 mm, flip angle (FA) = 90°, 30 diffusion-sensitive directions with b = 1, 000 mm²/s, 5 volumes with b = 0 mm²/s, and 75 slices without inter-slice gap. All the DTI data were scanned twice to improve the signal-to-noise ratio (SNR). In addition, we acquired high resolution brain structural images with a T1-weighted 3D Ax FSPGR BRAVO sequence (TR = 8.212 ms, TE = 3.22 ms, inversion time = 450 ms, FA = 12°, data matrix = 256 × 256, FOV = 256 mm × 256 mm, voxel size = 1 × 1 × 1 mm³, slice thickness = 1 mm, and 136 axial slices covering the whole brain).

3. Data processing

3.1. Image preprocessing

DTI data preprocessing was performed in FDT, a tool implemented in FSL (http://fsl.fmrib.ox.ac.uk/fsl). First, for each subject, the two repeated DTI datasets were concatenated and averaged. Second, non-diffusion-weighted images (b0) were extracted and non-brain tissues were removed using FSL/bet. Third, head-motion and eddy current-induced distortions were corrected using FSL/eddy_correct, and gradient orientations of the B-matrix were rotated (Leemans and Jones, 2009). Then, the eigenvalues of diffusion tensor were estimated using FSL/dtifit, and fractional anisotropy (FA), radial diffusivity (RD), axial diffusivity (AD), and mean diffusivity (MD) were generated. Finally, a 2-tensor model was fitted at each voxel for necessary
preparation for the probabilistic tractography using FSL/bedpostx.

The structural images were preprocessed using an automatic pipeline called “recon-all” in FreeSurfer (http://surfer.nmr.mgh.harvard.edu/). The pipeline involved the following steps: 1) motion correction and intensity normalization, 2) brain extraction, 3) segmentation of WM, gray matter (GM), and cerebrospinal fluid (CSF), and 4) WGB and pial surface generation.

3.2. TBSS analyses on the SWM

Fig. 1 illustrates the pipeline for TBSS analyses on the SWM. The steps were as follows: a) Seeds mask creation. Individual WGB mask was extracted from the WM/GM interface surface from the FreeSurfer and transformed to the diffusion space using the affine transformation (dof = 12) as the seed mask of each subject (Fig. 1a). b) Exclusion mask creation. FA images of all subjects were initially registered to the FMRIB58_FA template in the MNI space by using the TBSS non-linear transformation. Then the Johns Hopkins University ICBM-DTI-81 white-matter labels atlas (Wakana et al., 2007) was transformed into individual diffusion space using the inverse TBSS registration as the exclusion mask for each subject (Fig. 1b). c) SWM mask reconstruction. We generated probabilistic streamlines, which were sampled distributions of voxel-wise principal diffusion directions, using FSL/protrackx. After the probabilistic tractography, each individual tractographic result was normalized by dividing by the total number of the streamlines and then thresholded at 0.01% (Nazeri et al., 2015) to remove across-subjects variations. All the normalized SWM fibers were then non-linearly registered to MNI space using the TBSS non-linear transformation. We averaged the registered SWM masks across the subjects and created a probabilistic map for a population-based SWM map. Because of the inter-subject variability, we adopted a threshold of 50% (Nazeri et al., 2015) to create the SWM mask (Fig. 1c). d) TBSS analyses. The FA images for all subjects were aligned to the FMRIB58_FA template using the non-linear registration in FSL/fnirt. The non-linear transformation from individual diffusion space to MNI space was generated, and we inverted the transformation to create individual exclusion masks. Next, the registered FA images were averaged to create whole-brain mean FA images. To limit the TBSS on the SWM region, the mean FA images were multiplied by the SWM mask to create the mean $FA_{SWM}$ images, and a skeleton procedure was applied to the mean $FA_{SWM}$ images to reconstruct the $FA_{SWM}$ skeleton, which is representative of SWM tracts with a line of common center. To remove non-WM tissues and false-positive fibers from the probabilistic results, we applied a threshold of FA > 0.2 to the $FA_{SWM}$ skeleton. To provide more information about the diffusion properties of the SWM, we also considered the $MD_{SWM}$, $RD_{SWM}$, and $AD_{SWM}$ by projecting them onto the $FA_{SWM}$ skeleton using a non-linear transformation.
3.3. Statistics

3.3.1. Demographic statistics

All the demographic statistics were calculated using SPSS 19.0. We used two-sample t-tests to evaluate differences in age and education level between the two groups. For the gender factor, we used the χ²-test to calculate the group difference in gender. The statistical significance level for each test was set at \( p < 0.05 \) (two-tailed).

3.3.2. TBSS analyses

The calculation steps were as follows. (i) A general linear model (GLM) was applied to assess difference in FA_SWM between groups using a non-parametric permutation test (5000 permutations) (Winkler et al., 2014). In this study, because the demographic factors of the BD patients were not perfectly matched with those of the HC group, we took gender, age, and education level as covariates when testing the FA_SWM difference between groups. We also assessed the between-group differences in the mean diffusivity (MD_SWM), axial diffusivity (AD_SWM), and radial diffusivity (RD_SWM) in the SWM by using the statistical method same to that of the FA_SWM analysis. The significance level was set at \( p < 0.05 \), which was corrected by a family-wise error (FWE) rate method. The clusters, which were corrected by a threshold-free cluster enhancement (TFCE) method (Smith and Nichols, 2009), were reported according to the LNAO-SWM79 atlas (Guevara et al., 2017). The LNAO-SWM79 atlas was adopted to determine the SWM bundles using deterministic tractography based on Desikan-Killiany parcellation. From the overlapping regions between the clusters and the atlas, we could determine the SWM bundles and associated cortical regions that showing significant between-group differences in FA_SWM, MD_SWM, RD_SWM, and AD_SWM.

3.3.3. Relationship between diffusion properties and clinical variables

In order to detect the relationship between diffusion properties and clinical variables of the BD patients, we first extracted the mean values of the FA_SWM, MD_SWM, AD_SWM, and RD_SWM in the clusters showing significant differences between groups, and then calculated the correlation between each of them and disease exacerbations, onset age, duration time of the BD patients. In the calculations, we also took gender, age, and educational level as covariates, and set the significance level at \( p < 0.05 \).

4. Results

4.1. Demographics

Table 1 lists demographic factors for each group and the clinical performance of the BD patients in this study. No significant difference was observed in either gender or age between the BD patients and the HC group (\( p > .05 \)). A significant difference in education level was found between the two groups (\( p < 0.01 \)).

4.2. Permutation results for the SWM

TBSS analyses showed no significant difference in AD_SWM between the BD group and the HC group, but we found significant between-group differences in FA_SWM, RD_SWM, and MD_SWM (Table 2). We found that compared to the HC group, the BD patients exhibited significantly uniformly reduced FA_SWM in five clusters (Fig. 2), which were labelled as Cluster-5, Cluster-4, Cluster-3, Cluster-2, and Cluster-1 from the largest to the smallest volume. According to the LNAO-SWM79 atlas, we determined the associated cortical regions and SWM bundles of these clusters (Fig. 3). We found that Cluster-5 is mainly distributed in the left prefrontal regions and located in the PoC-PrC, Op-PrC, RMF-SF, LOF-RMF, CMF-Op, Op-SF, CMF-RMF, Tr-SF, RAC-SF, and CMF-PrC bundles. Cluster-4 is distributed in the left fronto-parietal regions and located in the PoC-PrC bundles. Cluster-3 is distributed in the right prefrontal regions and located in the RMF-SF, Op-SF, and Tr-SF bundles. Cluster-2 is distributed in the left parietal regions, and located in the IP-SF and SP-SM bundles. Cluster-1 is distributed in the right prefrontal regions, and located in the RMF-SF and RAC-SF bundles. We also found significantly increased RD_SWM and MD_SWM in the BD patients compared to the HC (Fig. 4). Similar to the labelling method of FA_SWM clusters, we found the two clusters with significant increased MD_SWM are distributed in the right frontal regions. The MD_SWM Cluster-1 is located in the CMF-PrC and Op-PrC bundles, and the MD_SWM Cluster-2 is located in the CMF-SF and CMF-PrC bundles. As for RD_SWM, the cluster with significantly increased RD_SWM is distributed in the right prefrontal cortex and is located in the RMF-SF and Tr-SF bundles.

4.3. Relationship between diffusion properties and clinical variables

The correlation analyses showed that there was no significant correlation between RD_SWM/MD_SWM and clinical variables of the BD patients. We found the FA_SWM between the viPFC and the dIPFC (Cluster-3) correlated negatively with the disease exacerbations of the BD patients (Fig. 5).

5. Discussion

This study focused on SWM changes in patients with BD. We used an automatic probabilistic tractography approach to restrict the fibers near the cortex and discard the fibers passing through the exclusion regions. In this study, the BD patients showed an abnormally reduced FA_SWM around nodes in the dIPFC, viPFC, mPFC, and parietal regions, indicating disrupted connections within the emotion regulatory network of the BD patients.

5.1. FA_SWM in the dIPFC

In the dIPFC, the BD patients showed reduced FA_SWM in the bilateral RMF-SF and the left CMF-RMF bundles compared to the HC group. The dIPFC is believed to be involved in working memory (Wager and Smith, 2003; Curtis, 2006), planning (Rosenbloom et al., 2012), and attention control (Hopfinger et al., 2000). In the emotion regulatory network, the dIPFC is assumed to be responsible for manipulating cognitive appraisals (Zilverstand et al., n.d.; Buhle et al., 2014). Previous studies (Frank et al., 2014; Eippert et al., 2007) have reported increased activation in the dIPFC, indicating disrupted connections within the emotion regulatory network of the BD patients. The abnormal activation in the dIPFC may be related to emotion dysregulation in BD patients. In this study, reduced FA_SWM in the bundles within the dIPFC, indicating disrupted cortico-cortical connections within the dIPFC, may reflect the abnormal emotion regulation of BD patients.

5.2. FA_SWM between the LOF and the dIPFC

In this study, the BD patients showed reduced FA_SWM in the left LOF-RMF bundle compared to the HC group. Previous studies suggested that LOF activation may be related to different responses to punishments (Kringlebach and Rolls, 2004) or to value expectations (Bermohl et al., 2010). For example, BD patients exhibited increased activation under an increasing gain expectation and decreased activation under an increasing loss expectation, but the controls had an inverse activation in the LOF (Bermohl et al., 2010). This inverse activation in the LOF between BD patients and controls under loss- or gain-expectation conditions may indicate that BD patients underestimate the punishments that may result from their behaviors. In this study, reduced FA_SWM in the bundles between the LOF and RMF, indicating altered connections around the LOF, may reflect abnormality in the LOF in BD patients.
5.3. FA\textsubscript{SWM} between the vlPFC and the dlPFC

In this study, compared to the HC group, the BD patients showed reduced FA\textsubscript{SWM} in the bilateral Op-SF, Tr-SF, and left CMF-Op bundles, passing from the dlPFC to the vlPFC. The vlPFC and the dlPFC are highly interconnected (Goulas et al., 2012) in the human brain, and the vlPFC was found to play an inhibitory role in cognitive, motor, and affective responses in the emotion regulation network (Berkman et al.,

### Table 2

Clusters showing significant differences between the patients with bipolar disorder (BD) and the healthy controls (HC).

| Cluster | MNI peak coordinates | Cluster size (mm\(^3\)) | Mean value | p-value \(^a\) | Hemi-sphere | Associated regions \(^b\) |
|---------|----------------------|--------------------------|------------|----------------|-------------|-------------------------|
| FA\textsubscript{SWM} | | | | | | |
| Cluster-1 | 19 | 49 | 17 | 41 | 0.40 | 0.43 | 0.04 | R | RMF, SF, RAC |
| Cluster-2 | -28 | -52 | 35 | 118 | 0.34 | 0.37 | 0.028 | L | IP, SP, SM |
| Cluster-3 | 19 | 25 | 36 | 158 | 0.49 | 0.52 | 0.002 | R | RMF, Op, Tr, SF |
| Cluster-4 | -34 | -27 | 44 | 176 | 0.42 | 0.44 | 0.028 | L | PoC, PrC |
| Cluster-5 | -19 | 13 | 43 | 1,173 | 0.42 | 0.44 | 0.012 | L | PoC, PrC, Op, Tr, RMF, CMF, LOF, SF, RAC |
| MD\textsubscript{SWM} | | | | | | | |
| Cluster-1 | 48 | -1 | 36 | 50 | 7.6 \times 10^{-4} | 7.3 \times 10^{-4} | 0.038 | R | CMF, Op, PrC |
| Cluster-2 | 36 | -2 | 40 | 60 | 7.4 \times 10^{-4} | 7.2 \times 10^{-4} | 0.045 | R | CMF, SF, PrC |
| RD\textsubscript{SWM} | | | | | | | |
| Cluster-1 | 20 | 50 | 11 | 86 | 5.9 \times 10^{-4} | 5.7 \times 10^{-4} | 0.035 | R | RMF, Tr, SF |

Abbreviations: RMF, rostral middle frontal gyrus; SF, superior frontal gyrus; RAC, rostral anterior cingulate gyrus; IP, inferior parietal gyrus; SP, superior parietal gyrus; SM, supramarginal gyrus; Op, pars opercularis gyrus; Tr, pars triangularis gyrus; PoC, postcentral gyrus; PrC, precentral gyrus; CMF, caudal middle frontal gyrus; LOF, lateral orbitofrontal gyrus; L (R): left (right)-hemisphere. All the abbreviations were defined based on the Desikan-Killiany atlas. FA\textsubscript{SWM}/MD\textsubscript{SWM}/RD\textsubscript{SWM}, fractional anisotropy/mean diffusivity/radial diffusivity in the superficial white matter.

\(^a\) The \(p\)-value of the cluster-based peak coordinates under \(FWE\) correction.

\(^b\) The cortical regions connected by SWM bundles with reduced FA\textsubscript{SWM} were obtained from the tract-based spatial analysis (TBSS). These regions were determined according to the LNAO-SWM\textsubscript{79} atlas.

5.3. FA\textsubscript{SWM} between the vlPFC and the dlPFC

In this study, compared to the HC group, the BD patients showed reduced FA\textsubscript{SWM} in the bilateral Op-SF, Tr-SF, and left CMF-Op bundles, passing from the dlPFC to the vlPFC. The vlPFC and the dlPFC are highly interconnected (Goulas et al., 2012) in the human brain, and the vlPFC was found to play an inhibitory role in cognitive, motor, and affective responses in the emotion regulation network (Berkman et al.,

**Fig. 2.** Presentations of the tract-based spatial statistics (TBSS) results in FA\textsubscript{SWM}. a) Surface-based representation. The five clusters showing reduced FA\textsubscript{SWM} in the patients with bipolar disorder (BD) compared to the controls were projected on an inflated surface. b) Volumetric representation. The five clusters were displayed with three-dimensional view at their peak MNI coordinates. L, left hemisphere; R, right hemisphere. FA\textsubscript{SWM}, fractional anisotropy in the superficial white matter.
Fig. 3. Presentations of associated cortical regions and superficial white matter (SWM) bundles with significantly reduced $FA_{SWM}$ in patients with bipolar disorder (BD) compared to controls. a) shows the associated cortical regions, which were connected by disrupted SWM bundles. b) shows the associated SWM bundles with significantly reduced $FA_{SWM}$. $FA_{SWM}$, fractional anisotropy in the superficial white matter. SF, superior frontal gyrus; RAC, rostral anterior cingulate gyrus; RMF, rostral middle frontal gyrus; CMF, caudal middle frontal gyrus; IP, inferior parietal gyrus; SP, superior parietal gyrus; SM, supramarginal gyrus; Op, pars opercularis gyrus; Tr, pars triangularis; PoC, postcentral gyrus; PrC, precentral gyrus; LOF, lateral orbitofrontal gyrus.

Fig. 4. Presentations of the tract-based spatial statistics (TBSS) results in $MD_{SWM}$ and $RD_{SWM}$. a) shows the cortical regions, which were connected by the disrupted SWM bundles. b) shows the SWM bundles with significantly changed $MD_{SWM}$ or $RD_{SWM}$ separately. c) displays the clusters showing significantly changed $RD_{SWM}$ or $MD_{SWM}$ with three-dimensional view at their peak MNI coordinates. L, left hemisphere; R, right hemisphere. $RD_{SWM}$/$MD_{SWM}$, radial diffusivity/mean diffusivity in the superficial white matter. PrC, precentral gyrus; CMF, caudal middle frontal gyrus; RMF, rostral middle frontal gyrus; Op, pars opercularis gyrus; SF, superior frontal gyrus; Tr, pars triangularis.
In this study, the BD patients showed reduced $FA_{SWM}$ in the bilateral RAC-SF bundles, which connect the rostral part of the anterior cingulate cortex (RAC) and the dIPFC, compared to the HC group. The RAC is believed to be involved in affective functioning (Bush et al., 2000). Matsuo et al. (Matsuo et al., 2009) found that the RAC volume was negatively correlated with impulsivity in BD patients. Several studies (Wagner et al., 2006; Matsuo et al., 2007) involving attention and working memory tasks have reported that patients with major depression disorder (MDD) showed increased activation in the RAC compared to controls, compensating for the dIPFC, to meet task demands. For example, Wagner et al. (Wagner et al., 2006) reported that during Stroop tasks MDD patients showed highly positive correlations between activity in the dIPFC, the RAC, and interference scores, but the controls showed no significant correlations between these. Treadway et al. (Treadway et al., 2009) suggested that this positive correlation in MDD patients may reflect cortical inefficiency between the RAC and the dIPFC. In this study, the reduced $FA_{SWM}$, showing a weak connection between the RAC and the dIPFC, may reflect cortical inefficiency between the RAC and the dIPFC in BD patients.

5.5. $FA_{SWM}$ in the fronto-parietal regions

In this study, the BD patients showed reduced $FA_{SWM}$ in the bundles in the left IP-SP, SP-SM, and PoC-PrC, compared to the HC group. Niendam et al. (Niendam et al., 2012) found that the parietal cortex is involved in the cognitive control network and that the IP and SP are activated as well as the dIPFC in inhibition or in working memory tasks. Recent meta-analyses (Zilverstand et al., n.d.; Buhle et al., 2014) revealed that the emotion regulatory network largely overlaps with the control network, in which the IP and SP are involved in allocating attention and salience detection (Corbetta and Shulman, 2002). Picó-Pérez et al. (Picó-Pérez et al., 2017) reported an aberrant activation in mood and anxiety disorder patients, who displayed increased activation in the SM, SP, and PrC but reduced activation in the vlPFC, mPFC, and posterior cingulate cortex (PCC) compared to a control group while regulating negative emotions. This increased activation in the SM, SP, and PrC may indicate a compensatory mechanism for impaired cortical control of negative emotions in the patients. In this study, reduced $FA_{SWM}$ around the IP, SP, and SM, showing the abnormal connections around them, may reflect abnormality in nodes of the cognitive control network in BD patients.

In addition, we found reduced $FA_{SWM}$ in the PoC-PrC, Op-PrC, and CMF-PrC bundles in the left hemisphere. Although the PoC and PrC are in the somatosensory and motor region, they have been found to be important in discriminating emotions (Saarimäki et al., 2016). Kujawa et al. (Kujawa et al., 2015) found that activation in the PrC and PoC when responding to threatening faces better predicted the treatment response in adult anxiety disorder patients, suggesting an important role for the PrC and PoC in emotion regulation. In this study, reduced $FA_{SWM}$, showing abnormal connections around the PoC and PrC, may reflect abnormality in PrC and PoC and was related to disrupted emotion regulation of BD patients.

5.6. Relationship between $FA_{SWM}$ and disease exacerbations

This study showed that $FA_{SWM}$ of the Cluster-3 exhibited negative correlation with disease exacerbations of the BD patients. Cluster-3 comprised the right RMF-SF, Op-SF, and Ts-SF bundles passing from the vlPFC to the dIPFC. Townsend et al. (Townsend and Altshuler, 2012) indicated that dysfunction in the vlPFC may result in emotion dysregulation and make BD patients susceptible to relapse into the manic state. The connection between the dIPFC and vlPFC was found to mediate the emotion regulatory process (Morawetz et al., 2016). Thus, reduced $FA_{SWM}$ indicating a weak connection between the vlPFC and the dIPFC, may influence the emotion regulatory process and make BD patients more susceptible to the manic state.

5.7. Limitations

Several limitations in this study should be addressed. First, the main limitation arises from the sample size of the BD patients ($n = 37$), which may preclude the validation of its inferences with respect to other clinical populations with BD. Second, specific scales for measuring the emotion regulatory ability of BD patients were not used; we only utilized the HAMD and YMRS scales. Thus, the relationship between the emotion regulatory ability of BD patients and diffusion properties in the SWM could not be detected in this study. Third, this study primarily analyzed DTI data without a functional MRI data analysis, so we could not provide study-specific relationships between brain structure and function. Further SWM studies in BD patients should consider using multi-modal data. Fourth, the goal of this study was to detect SWM differences in the BD patients compared to the HC, so we did not consider the subcortical regions, which are known to be very important in the emotion regulatory network.

6. Conclusion

In conclusion, this study presents altered cortico-cortical connections near the regions related to emotion dysregulation in BD patients. The findings indicated that the SWM may serve as the brain structural basis underlying the disrupted emotion regulation of BD patients. In contrast with the emotion regulatory network, the disrupted $FA_{SWM}$ in the parietal cortex may indicate that emotion dysregulation in BD patients is related to the cognitive control network.

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Competing interests statement

The authors declare that they have no competing financial interests.

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