

Widely Impaired White Matter Integrity and Altered Structural Brain Networks in Psychogenic Non-Epileptic Seizures

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Objective: The underlying neural correlates of psychogenic non-epileptic seizures (PNES) are still unknown and their identification would be helpful for clinicians and patients. This study aimed to reveal details of white matter microstructure and alterations in brain structural networks in patients with PNES by using diffusion tensor imaging (DTI) and graph theoretical connectivity analysis.

Methods: Seventeen patients with PNES and 26 age- and sex-matched healthy controls were enrolled. All participants underwent DTI on a 3.0-T MRI scanner, and fractional anisotropy (FA) and mean diffusivity (MD) maps were compared by tract-based spatial statistics. Additionally, the structural networks derived from DTI data were analyzed using graph theory and two different parcellation schemes.

Results: Patients with PNES showed widespread decreases in FA and increases in MD, particularly in the deep white matter. In addition, graph theoretical analysis revealed impaired brain networks in PNES, including increased path length, decreased network efficiency, altered nodal topology, and reduced regional connectivity in the right posterior areas.

Conclusion: We found widely impaired white matter integrity and impaired brain structural networks in Japanese patients with PNES. These findings contribute to the accumulation of evidence on PNES and may improve understanding of this condition.

Keywords: diffusion tensor imaging, brain structural networks, graph theory, psychogenic non-epileptic seizures, magnetic resonance imaging

Introduction

Psychogenic non-epileptic seizures (PNES) are medically unexplained seizure-like episodes that are associated with psychological processes. They are a common differential diagnosis of epilepsy and a major health problem around the world.^{1,2} While the underlying mechanism of PNES would be different from that of epilepsy, improved neurobiological knowledge of PNES will be helpful for both treatment and patients' understanding of their own illness.^{3,4}

Although neuroimaging studies are expected to provide relevant information and insights about the mechanisms, there are still a limited number of papers.⁴ Additionally, the social and medical environments of PNES vary widely around the world¹ and, similarly, the potential causes of PNES, such as abuse or family conflict, are also dependent on background cultures.² Thus, further accumulation of evidence on neurobiological aspects of PNES from various cultures may contribute to better understanding and clinical practice in this field.

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We recently reported an abnormal brain aging process in Japanese patients with PNES based on T1-weighted morphological neuroimaging.⁵ On the other hand, diffusion tensor imaging (DTI) has conventionally been used to evaluate white matter tracts through microstructural parameters such as fractional anisotropy (FA) and mean diffusivity (MD),⁶ which might provide additional information on the brain in PNES. Recent technical advances have also enabled us to evaluate structural connectivity and brain networks.⁷ Given that network analysis has provided further insights into brain abnormalities in many neurological and psychiatric disorders,⁷ PNES may also benefit from this approach by revealing potential brain mechanisms and/or predispositions to psychogenic seizures.

Accordingly, this study used DTI and graph theoretical connectivity analysis to explore details of white matter microstructure and alterations in brain structural networks in patients with PNES.

Materials and Methods

Participants

We retrospectively enrolled consecutive patients with PNES at our institute between January 2012 and December 2017, based on the following inclusion criteria: (1) presence of PNES confirmed on video-EEG monitoring, (2) no epileptiform discharges or epileptic seizures on video-EEG monitoring, and (3) no visual abnormality on MRI. Patients with the following criteria were excluded: (1) a significant past medical history of acute encephalitis, meningitis, severe head trauma, or ischemic encephalopathy; (2) low intelligence quotient (i.e. <69) in Wechsler Adult Intelligence Scale III. Seventeen patients were enrolled, consequently.

For healthy controls, we also recruited 26 age- and sex-matched volunteers with no history of neurological or psychiatric disease and no use of medication affecting the central nervous system. We found no possible structural anomalies or abnormalities in the controls on MRI. The detailed demographics of both groups are shown in Table 1.

This retrospective study was approved by the institutional review board at National Center of Neurology and Psychiatry Hospital (A2017-060), and the need for patient informed consent was waived.

MRI Acquisition

The MRI scans for all participants were performed on a 3.0-T MR system with a 32-channel coil (Philips Medical Systems, Best, The Netherlands). The parameters of the DTI

Table 1 Demographics of Healthy Controls and Patients with PNES

| Feature | PNES (n=17) | HC (n=26) | p-value |
|---|-------------------------------|-------------------------------|-------------------|
| Sex (n) Male:female | 4:13 | 6:20 | 0.97* |
| Age at Examination (Years) Mean \pm SD [95% CI] | 28.7 \pm 9.4 [23.9–33.5] | 29.6 \pm 6.9 [26.8–32.4] | 0.72 [†] |
| Disease Duration (Years) Mean onset age \pm SD [95% CI] | 22.4 \pm 9.8 [17.4–27.5] | N/A | |
| Mean duration of PNES \pm SD [95% CI] | 6.3 \pm 5.9 [3.3–9.3] | N/A | |

Notes: *Pearson's χ^2 test, [†]two-sample t-test.

Abbreviations: PNES, psychogenic non-epileptic seizure; HC, healthy control; CI, confidence interval; N/A, not available.

sequence were as follows: repetition time (TR)/echo time (TE), 6700 ms/58 ms; flip angle, 90°; number of excitations (NEX), 2; slice thickness, 3.0 mm with no gap; slices, 60; matrix, 80 \times 78; and field of view (FOV), 24 \times 24 cm. Diffusion was measured along 15 non-collinear directions using a diffusion-weighted factor b of 1000 s/mm², and one image was acquired with no diffusion gradient.

We also performed MRI examinations with the following protocols: (1) for three-dimensional (3D) sagittal T1-weighted images: TR/TE, 7.12 ms/3.4 ms; flip angle, 10°; NEX, 1; slice thickness, 0.6 mm with no gap; slices, 300; matrix, 260 \times 320; and FOV, 26 \times 24 cm; (2) for transverse turbo spin echo T2-weighted imaging: TR/TE, 4704/80 ms; flip angle, 90°; NEX, 2; thickness, 3.0 mm with a 1.5-mm gap; slices, 34; matrix, 368 \times 215; and FOV, 23 \times 18 cm; and (3) for coronal fluid-attenuated inversion recovery (FLAIR) imaging: TR/TE, 10,000/120 ms; inversion time, 2450 ms; flip angle, 120°; NEX, 2; thickness, 3.0 mm with a 1.5-mm gap; slices, 34; matrix, 272 \times 144; and FOV, 23 \times 18 cm.

Tract-Based Spatial Statistics

The DTI data were initially processed with PANDA toolbox v.1.3.1 (<https://www.nitrc.org/projects/panda/>)⁸ running with FMRIB Software Library (FSL) version 5.0.11 and MATLAB 2018b. After eddy current correction and brain extraction, the software yielded voxel-wise maps of FA and MD for each participant. To compare FA and MD maps between PNES and healthy controls, we performed tract-based spatial statistics (TBSS).⁶ Skeletonized FA and

MD data were analyzed to evaluate the differences between the two groups with age and sex as covariates using the Threshold-Free Cluster Enhancement option in FSL Randomize with 10,000 permutations.

Network Analysis

Individuals' structural white matter networks were constructed using the PANDA toolbox with the number of fibers as deterministic, weighted edges derived from DTI data. All the tracts were computed by seeding each voxel with an FA that was > 0.2 . Additionally, the 3D-T1 images were used to define the network nodes. Each FA image was co-registered to its corresponding 3D-T1 image by affine transformation. Then, the 3D-T1 image was non-linearly registered to the ICBM152 template in the MNI space. Using inverse transformation, the standard gray matter atlases were warped to each participant's native space and used for node definition. The detailed pipelines for network constructions are described in the original PANDA toolbox publication.⁸

To enhance validity and reproducibility, we adopted two different parcellation schemes for graph theoretical analysis: the Automated Anatomical Labeling (AAL) atlas with 90 regions of interest (ROIs)⁹ and the Harvard-Oxford atlas (HOA) with 110 ROIs.^{10,11} The names and locations of the ROIs are available in [Supplementary Tables 1 and 2](#).

We performed graph theoretical analysis using the GRETNA toolbox (<https://www.nitrc.org/projects/gretna/>).¹² In this study, we calculated global network properties, including the small-worldness property (clustering coefficient and characteristic path length) and global network efficiency (global efficiency and local efficiency), as well as nodal properties (nodal betweenness centrality). Nodal

betweenness centrality is calculated as the number of shortest paths between any two nodes that pass through the node and represents the importance of the node.¹³ The detailed definitions and meanings of each network property are described elsewhere.¹³ In addition, a network-based statistic (NBS) method was used to identify edges with significantly different connectivity strengths between the two groups.

Statistics

For TBSS, $p < 0.05$ with family-wise error (FWE) correction by FSL software was deemed significant. For comparison of networks, statistical analyses were performed with the GRETNA toolbox. Between the two groups, global network properties were compared by two-sample *t*-test, whereas the nodal betweenness centrality of each node was compared by two-sample *t*-test with Bonferroni correction. Additionally, NBS correction for multiple comparisons was used to compare connectivity strength. In all statistics for networks, two-sided $p < 0.05$ was considered significant.

Results

The results of the TBSS comparison are shown in [Figure 1](#). We found widely decreased FA and increased MD in patients with PNES. In particular, impairment was found mainly in the deep white matter.

As for the global network properties ([Figure 2](#) and [Table 2](#)), there was a significantly increased characteristic path length and decreased global and local brain network efficiency in PNES with both parcellation schemes, namely, AAL and HOA.

[Figure 3](#) shows the regional nodal betweenness centrality of each group as well as the statistical differences

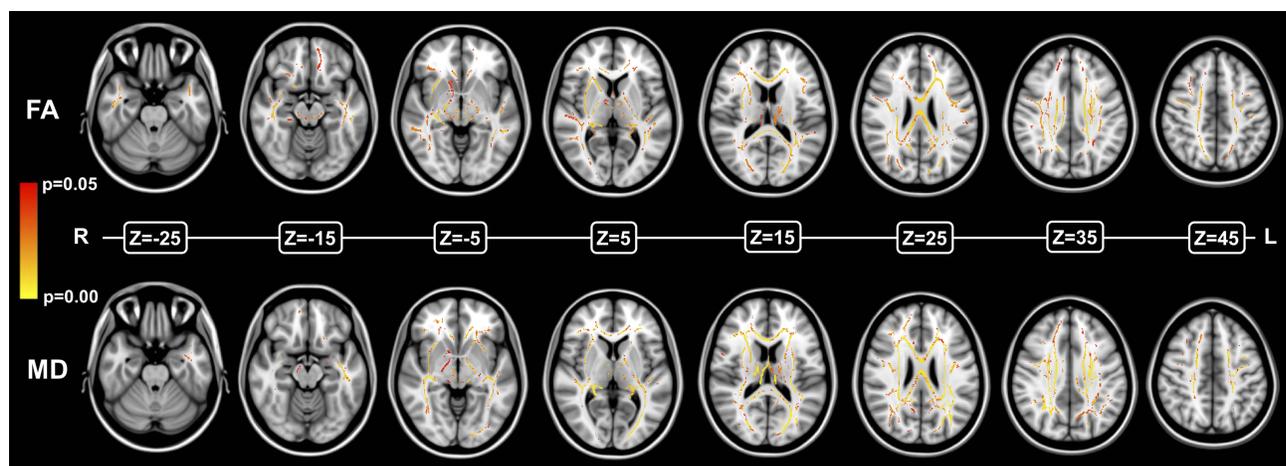


Figure 1 Compared with healthy controls, patients with PNES showed widely decreased FA (upper) and increased MD (lower) on TBSS comparison.

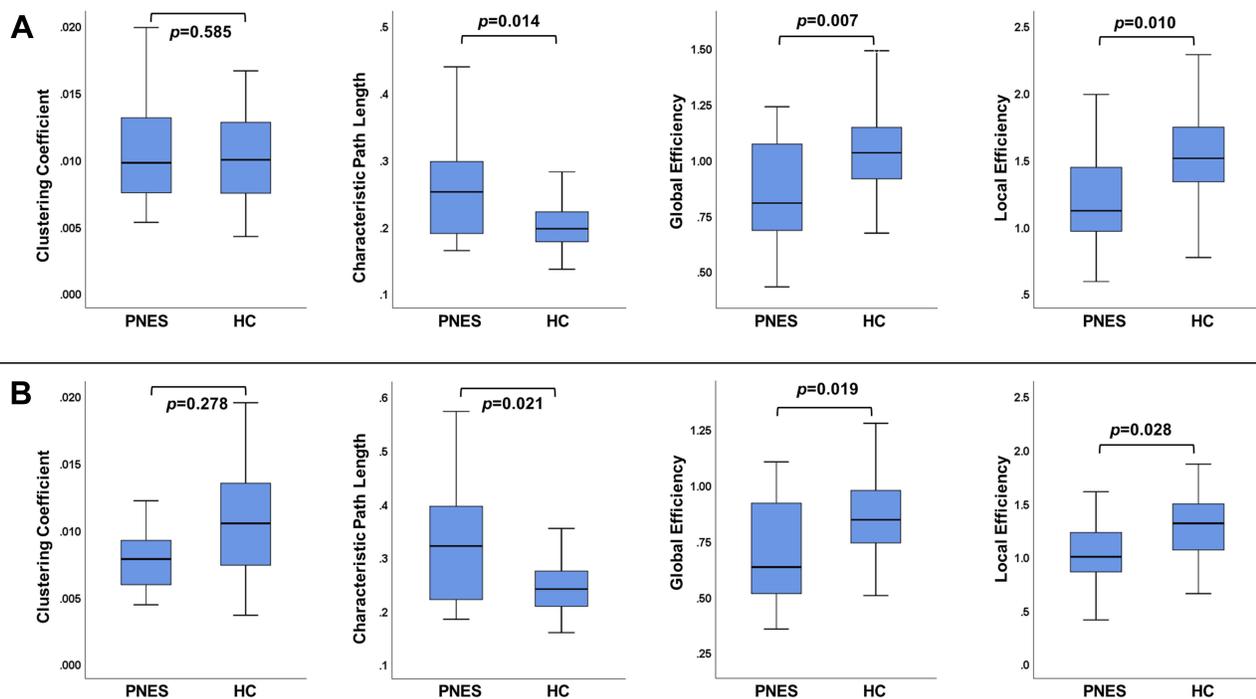


Figure 2 Results of comparisons of global network properties with the AAL atlas (A) and the HOA atlas (B).

between the two groups. The results suggested that nodal properties may be subject to atlas differences.

Finally, we found a significantly reduced connectivity network in the right posterior brain with the NBS analysis and AAL parcellation scheme (Figure 4), whereas there was no significance with HOA.

Discussion

In this study, we used DTI and graph theory to investigate white matter microstructures and brain networks in Japanese patients with PNES. We found widespread FA

and MD abnormalities as well as impaired brain networks including increased path length, decreased network efficiency, altered nodal topology, and reduced network connectivity. Considering the need for additional evidence from neuroimaging,⁴ our findings may contribute to the neurobiological understanding of PNES.

A few studies have investigated PNES using DTI.^{14–16} Hernando et al¹⁵ focused on the uncinate fasciculus and found significant asymmetry of the uncinate fasciculus in patients with PNES. The same group also reported increased FA in the left hemisphere in PNES using whole-brain

Table 2 Comparison of Global Metrics Between Healthy Controls and Patients with PNES

| Feature | PNES [Mean ± SD] | HC [Mean ± SD] | p-value | Mean Difference [95% CI] | Effect Size [Cohen's d] |
|--|---------------------|-------------------|---------|--------------------------|-------------------------|
| Metrics from AAL Parcellation | | | | | |
| Clustering coefficient [$\times 10^2$] | 1.04 ± 0.38 | 0.98 ± 0.32 | 0.585 | 0.06 [−0.16–0.29] | 0.17 |
| Characteristic path length | 0.27 ± 0.09 | 0.20 ± 0.04 | 0.014 | 0.06 [0.01–0.12] | 0.89 |
| Global efficiency | 0.84 ± 0.26 | 1.05 ± 0.22 | 0.007 | 0.21 [0.06–0.36] | 0.87 |
| Local efficiency | 1.19 ± 0.40 | 1.52 ± 0.38 | 0.010 | 0.33 [0.08–0.58] | 0.85 |
| Metrics from HOA Parcellation | | | | | |
| Clustering coefficient [$\times 10^2$] | 0.90 ± 0.41 | 1.04 ± 0.44 | 0.278 | 0.15 [−0.12–0.42] | 0.35 |
| Characteristic path length | 0.33 ± 0.12 | 0.25 ± 0.06 | 0.021 | 0.08 [0.01–0.14] | 0.83 |
| Global efficiency | 0.70 ± 0.24 | 0.86 ± 0.18 | 0.019 | 0.16 [0.03–0.29] | 0.74 |
| Local efficiency | 1.05 ± 0.38 | 1.31 ± 0.36 | 0.028 | 0.26 [0.03–0.50] | 0.71 |

Abbreviations: PNES, psychogenic non-epileptic seizure; HC, healthy control; CI, confidence interval; N/A, not available.

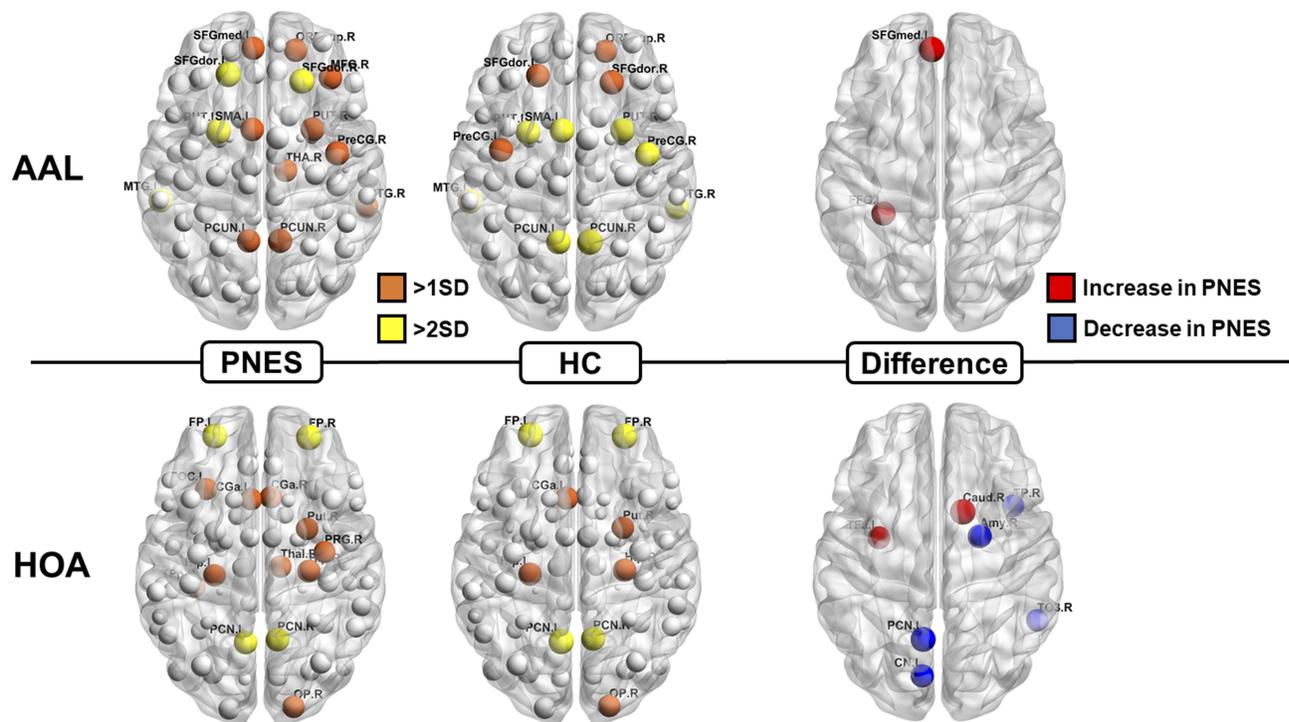


Figure 3 Locations of nodal betweenness centrality in the two groups (left and center) and significantly different nodes (right). The size of the node denotes the betweenness centrality in each group (left and center) and the t-values of the differences (right). The node abbreviations correspond to those of [Supplementary Tables 1 and 2](#).

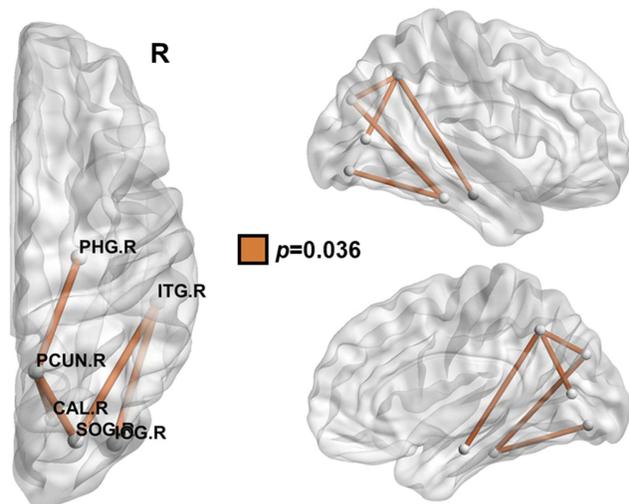


Figure 4 Compared with healthy controls, patients with PNES showed a reduced connectivity network in the right posterior areas, detected by NBS and the AAL atlas parcellation scheme. The node abbreviations correspond to those of [Supplementary Table 1](#).

TBSS.¹⁶ Their results seem to be different from ours, which might be attributable to patient selection or the underlying causes of PNES. Most studies on DTI findings in various neurological and psychiatric disorders^{17–22} showed reduced FA and/or increased MD, which were seen in our results, whereas increased FA are rarely reported.²³ The relationship

and interpretation between brain dysfunction and DTI parameters are complex, and the discrepancy in FA results in PNES might reflect the heterogeneity and complexity of this disorder. Another possible explanation for this divergence might be treatment differences (e.g., inappropriate long-term use of anti-epileptic drugs), which could vary among countries^{1,2} and affect brain structures.

Additionally, Ding et al¹⁴ using DTI and functional MRI reported altered functional and structural networks in PNES. They found not only increased path length, which is consistent with our results, but also an increased clustering coefficient. On the other hand, we have performed validation analysis using two different parcellation schemes, as well as NBS analysis, which may provide further evidence of impaired network properties in PNES. Using the NBS analysis, we found reduced connectivity in right posterior areas, involving the right precuneus and parahippocampal cortex, in PNES, although this difference was found in the AAL parcellation only. This finding may suggest some specific impairment of this region in patients with PNES, such as default mode network (DMN) dysfunction and alertness.²⁴

In fact, DMN abnormalities have been commonly reported in depression,²⁵ anxiety²⁶ and post-traumatic stress disorder (PTSD).²⁷ Given the close associations

with PNES,² we should bear in mind the effect of these psychiatric disorders as the confounding factors. As for the white matter structures, major depression showed increased path length,²⁵ which was also seen in the present study, whereas decreased path length is reported in PTSD.²⁸ The relationship between psychiatric disorders and brain connectome findings is complicated, and thus further investigations with detailed psychiatric symptoms will be needed to clarify the neurobiological complexities of PNES.

Compared with DTI studies, there have been a greater number of morphological statistical neuroimaging studies of PNES.^{29–33} In general, PNES shows frontal and temporal lobar abnormalities in gray matter volume or cortical thickness, although there are some inconsistencies. These areas are also related to normal aging during young adulthood to middle age,³⁴ which may explain the faster brain aging in patients with PNES.⁵ Interestingly, FA in the corpus callosum is the area most associated with the aging process.³⁵ Because we found a severe FA reduction mainly in the deep white matter, including the corpus callosum (Figure 1), these FA changes may indicate a strong link between PNES and abnormal brain aging.

The main limitations of this study are the small sample size and the lack of detailed neuropsychological and/or psychiatric information, which may make this study preliminary in nature. Details of demographics, such as handedness, family history of neurological diseases, and history of febrile seizures, were also unavailable. In particular, as stated previously, the lack of information on the comorbid psychiatric symptoms should be noted, because they are important confounding factors to interpret the present results. On the other hand, the number of neurobiological studies on PNES is still low, and this is the first neuroimaging study to investigate Japanese patients with PNES. In fact, we found several statistically significant findings in this preliminary cohort, which would help to guide further research into PNES around the world. In addition, to fully reveal the effects of PNES, another group of controls (e.g., psychiatric patients without PNES) might have been valuable, as suggested in a previous paper.⁴ Thus, more comprehensive and multicohort studies with larger sample sizes are desirable. Additionally, more high-precision imaging protocols should be needed, considering the relatively low number of DTI directions in this study (i.e. 15). Finally, the effects of drugs (e.g., psychoactive drugs or past use of anti-epileptic drugs) also need to be considered when the data from this study are being interpreted.

Conclusions

We found widespread decreases in FA and increases in MD, as well as impaired brain structural networks including increased path length, decreased network efficiency, and regional reduced connectivity in the right posterior region in Japanese patients with PNES. These findings contribute to the accumulation of evidence on PNES and may improve the understanding of this condition.

Ethics Approval and Informed Consent

In compliance with the Helsinki Declaration, this retrospective study using anonymized data was approved by the institutional review board at National Center of Neurology and Psychiatry Hospital (A2017-060), and the need for patient informed consent was waived.

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Disclosure

The authors report no conflicts of interest in this work.

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