ORIGINAL ARTICLE



Cerebellar Atrophy in Multiple System Atrophy (Cerebellar Type) and Its Implication for Network Connectivity

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Abstract

We sought to assess structural and functional patterns of cerebellum in multiple system atrophy (cerebellar type), and investigate the associations of structural and functional cerebellar gray matter abnormalities. We collected magnetic resonance imaging data of 18 patients with multiple system atrophy (cerebellar type) and 18 health control subjects. The gray matter loss across the motor and cognitive cerebellar territories in patients was assessed using voxel-based morphometry. And change in the connectivity between the cerebellum and large-scale cortical networks was assessed using resting-state functional MRI analysis. Furthermore, we assessed the relationship between the extent of cerebellar atrophy and reduced-activation in the cerebellar-cortical and subthalamo-cerebellar functional connectivities. We confirmed the gray matter loss across the motor and cognitive cerebellar territories in patients and found that the extent of cerebellar atrophy was correlated with decreased connectivity between the cerebellum and large-scale cortical networks, including the default, frontal parietal, and sensorimotor networks. The volume reduction in the motor cerebellar functional connectivity without the subthalamic nucleus atrophy. These results emphasized significant atrophy in the cerebellar subsystem and its association with the large-scale cortical networks in multiple system atrophy in the cerebellar subsystem and its association with the large-scale cortical networks in multiple system atrophy in the neural pathophysiology mechanisms of disease.

Keywords Multiple system atrophy \cdot Cerebellar atrophy \cdot Magnetic resonance imaging \cdot Functional connectivity \cdot Voxel-based morphometry

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Introduction

Multiple system atrophy (MSA) is a progressive neurodegenerative disease pathologically characterized by alpha synuclein-positive glial cytoplasmic inclusions (GCIs), which lead to the degeneration and death of neuron mainly in the striatum, cerebellum, and olivopontine structures [1]. Currently, MSA is mainly classified into a parkinsonian (MSA-p) and a cerebellar variant (MSA-c) [2]. In this study, we focused on MSA-c, which emphasizes on the cerebellum atrophy and dysfunction.

As a structure of the cerebello-thalamo-cortical circuit (CTC), the cerebellum is shown to exert a far-reaching influence on motor, cognitive and emotional functions, through its anatomical and functional interconnections with the basal ganglia and much of the cortical mantle [3, 4]. These heterogeneous functions have been suggested to indicate the existence of a "motor" cerebellum connecting to cortical motor regions, and a "cognitive" cerebellum connecting to prefrontal



and parietal cortices [3, 5–7]. And the CTC circuit procession [6, 8, 9] suggested that the anatomically distinct cerebellar subregions are projected to unique cortical targets, which confirmed the heterogeneous structure and function of the cerebellum. From a network perspective, the cerebellar motor subsystem is consistently associated with the sensorimotor network, while the cognitive subsystem is functionally connected with large-scale cognitive-related networks, including the executive control, salience, and default mode networks (DMN) [10, 11]. Although cerebellar architecture is arguably more complex than two distinct subsystems [10], the division into motor and cognitive cerebellar subsystems might provide an important framework for understanding how alterations of the cerebellum's structure and function in disease states.

Most neuroimaging studies focusing on the cerebellar atrophy and dysfunction patterns revealed changes in volume and functional connectivity pattern in the cerebellum, reflecting the pathological changes of MSA-c [12–14]. In a previous study, the disrupted motor network was reported associated with the cerebellum dysfunction in patients with MSA-c [15]. Recently, one study of MSA-c indicated the altered functional connectivity between cerebellum and large-scale cortical networks [16]. These findings provided evidences for the hypothesis of "disconnection syndrome"[17]. However, no previous studies reported the structural and functional patterns of motor and cognitive cerebellar subsystems in patients with MSA-c. And the association between the structure and function of the cerebellar subsystems in MSA-c is unclear.

Here, we hypothesized that there were different disrupted structural and functional patterns of cerebellum subsystems in patients with MSA-c; in addition, the extent of gray matter loss would be associated with alterations in cerebellarcortical resting-state connectivity. To realize this goal, we compared the cerebellar subregions atrophy between patients with MSA-c and health controls (HCs). And we investigated whether local volume reduction of cerebellar subsystems was associated with alterations in cerebellar-cortical functional connectivity. Furthermore, a post hoc analysis was conducted to determine subthalamic nucleus (STN) atrophy and the relationship between the cerebellar subgroups and STN, to identify whether the connectivity between cerebellar atrophic subregions and cerebellar input source might be affected. Finally, we explore the relationship between motor severity, connectivity, and atrophy of cerebellum in the patients.

Material and Methods

Participants

A total of 36 subjects were recruited from Dongfang Hospital of Beijing University of Chinese Medicine, including 18 patients with MSA-c and 18 healthy HCs (matched by age and educational level).

All subjects provided their written informed consent before being involved in the study. The research ethics committee of Dongfang Hospital of Beijing University of Chinese Medicine approved the study protocols.

The diagnosis of MSA fulfilled the criteria of probable MSA-c type defined by the American Academy of Neurology and the American Autonomic Society [2]. The inclusion criteria for HCs were as follows:

- 1. Absence of neurological or psychiatric disorders including stroke, depression, or epilepsy.
- 2. Absence of neurological deficiencies including visual or hearing loss.
- Absence of abnormal findings of brain MRI, such as infarction or focal lesions.

The subjects of hemorrhage, infarction, tumors, trauma, or severe white matter hyperintensity were excluded from the study. All individuals were evaluated by performing complete physical and neuropsychological examinations including mini-mental state examination (MMSE), Montreal Cognitive Assessment (MoCA), and Unified Multiple System Atrophy Rating Scale (UMSARS). The clinical examinations were performed on the day before undergoing functional magnetic resonance imaging (fMRI) scan.

MRI Acquisition and Analysis

Data Acquisition

MRI data acquisition was performed on a GE 3.0 T Discovery 750 scanner. Foam padding and headphones were used to control head motion and scanner noise. The data scan parameters of resting-state fMRI were as follows: repetition time (TR)/echo time (TE)/flip angle (FA) = 2000 ms/30 ms/90°, field of view (FOV) = 24×24 cm2, resolution = 64×64 matrix, slices = 36, thickness = 3 mm, gap = 1 mm, voxel size = $3.75 \times 3.75 \times 3$ mm³, and bandwidth = 2232 Hz/pixel. For registration purposes, high-resolution anatomical images were collected using a 3D brain volume (BRAVO) T1-weighted sequence with the following parameters: TR/TE/inversion time (TI)/FA = 8150 ms/3.17 ms/450 ms/12°, resolution = 256×256 matrix, slices = 188, thickness = 1 mm, voxel size = $1 \times 1 \times 1$ mm³.

Voxel-Based Morphometry Analysis

VBM analysis was performed using the Statistical Parametric Mapping software (SPM12;http://www.fil.ion.ucl.ac.uk/spm/software/spm12). Structural images of gray matter (GM), white matter and cerebrospinal fluid were segmented [18].

The initial affine of the GM concentration map was registered into the Montreal Institute of Neurology (MNI) space (http://www.mni.mcgill.ca/). Nonlinear warping of GM concentration images was performed using differential analytic registration (DARTEL) techniques and was resliced to a resolution of 1.5 mm³[19, 20]. To obtain the relative GM of each voxel, the GM concentration map was multiplied by a nonlinear determinant derived from spatial normalization. Finally, the GM images were smoothed using a Gaussian kernel of 8 mm × 8 mm × 8 mm full-width at half-maximum (FWHM) to compensate for residual between-subject anatomical differences. After spatial preprocessing, statistical analysis was performed using a smoothed normalized GM map.

The cerebellar subregions (CBMc and CBMm, respectively) were defined by using a validated probabilistic atlas of the human cerebellum available in the SUIT [21]. Bilateral Crus I and Crus II were defined as CBMc, and bilateral leaflets V, VI, VIIb, VIIIa, and VIIIb were defined as CBMm [3] (see Fig. 1). Extracting regional GM intensity measurements in contrast to the quiescent state pattern, creating the cognitive and motor cerebellar regions of interest CBMc and CBMm. The mean gray matter intensity value of non-zero voxels in the region of interest was extracted and averaged for each subject. The result represented value of cerebellar atrophy degree for each individual in the study. A voxel-wise general linear model (GLM) and family-wise error (FWE) correction were used for statistical analysis to compare GM intensity of cerebellar subregions between MSA patients and HCs. This analysis was also used in the published subthalamic nucleus (STN) defined by the middle-aged data set [22]. All statistical tests were two-sided, and a P value < 0.05 was considered statistically significant.

Resting-State Functional Connectivity Analysis

TheData Processing Assistant for Resting-State fMRI (DPARSF;http://rfmri.org/DPARSF) was used for

preprocessing functional images. Briefly, preprocessing included data conversion, the first 10 volumes removal, slice timing correction, and head motion correction. To spatially normalize the fMRI data, the realigned volumes were spatially standardized into the MNI space using the EPI template. The functional images were resampled into a voxel size of $3 \times 3 \times 3$ mm³. Then, the functional images were smoothed with a Gaussian kernel of 4 mm FWHM.

The putative resting-state network analysis was performed by using the group independent component analysis (ICA) of the GIFT software (http://icatb.sourceforge.net, version 1.3i). The number of individual components (ICs) was determined by using the minimum component length (MDL) standard [23]. The fMRI data of all participants was collected in a group, and the time dimension of the aggregated data set was reduced by principal component analysis (PCA) and estimated by IC using the Informax algorithm (with time course and spatial maps) [24, 25]. We selected the highly stable ICASSO toolbox [26], categorized it using a visual inspection of the average component map and its average power spectrum, by running Infomax 100 times and randomized and bootstrapped datasets under different initial conditions. Back-reconstructed for ICs and time courses of each participant were analyzed by using CICA3 [27]. The individuallevel components were obtained from back-reconstruction and were converted into a z-score. The z-score reflected the degree of time series of a given voxel correlates with the mean time series of its belonging component. We mainly focused on 6 large-scale cortical networks [28]: anterior default network (aDMN), posterior default network (pDMN), left frontal parietal network (IFPN), right frontal parietal network (rFPN), dorsal attention network (dAN), sensorimotor network (SMN). Statistical analysis of individual ICA component maps of all subjects was performed by a random-effect single-sample t test, creating the sample-specific large-scale network masks (P < 0.05, FWE correction, cluster size > 100).



Fig. 1 Seed region in the cerebellum. Sagittal (a, d), coronal (b, e) and axial (c, f) view of the CBMc (red) and CBMm (blue) ROIs superimposed to the spatially unbiased atlas template of the cerebellum and brainstem

Finally, the processed images were imported into the functional connectivity toolbox in DPABI (http://rfmri.org/dpabi). Several covariates, including parameters of head motion, white matter, cerebrospinal fluid (CSF) signal, and global average signal were regressed by multiple linear regression analysis. After preprocessing, time band-pass filtering (0.01-0. 08 Hz) of the fMRI data reduced the effects of a lowfrequency drift and high-frequency physiological noise, such as respiratory and cardiac rhythms. We extracted blood oxygen level-dependent time processes from each of the nuisancecorrected source regions of interest and the similar processed the STN. And then, we analyzed the strength of each pair-wise connection across the regions of interest within network using the Pearson correlation. We normalized these values by using Fisher's r-to-z transformation and then compared the differences of the values between HCs and patients by using Bonferroni-Holm correction for multiple comparisons^[29].

Relationship Between Cerebellar Gray Matter Intensity and Resting-State Functional Connectivity

We explored the relationship between cerebellar GM atrophy and resting-state functional connectivity within network and the STN.

The correlation between the cerebellar subregions atrophy and cerebello-cortical connectivity in HCs was used as a baseline; the scores of patients group were directly compared with HCs.

The positive differences between patients and HCs implied a relative increase in connectivity accompanied with cerebellar atrophy, which might reflect a compensatory mechanism, while the negative differences between the patients and the HCs indicated a decreased functional cerebellum-cortical connectivity with the cerebellar subregions atrophy.

Relationship Between Motor Severity, Connectivity, and Cerebellar Atrophy

To determine whether a relationship existed between motor severity, atrophy, and connectivity, we analyzed the relationship between each cerebellar subregion (CBMm and CBMc) atrophy scores and the motor function as measured by UMSARS II by using a Pearson's correlation. The relationships between the functional connectivity changes of each network and the UMSARS II were also examined [30].

Statistical Analysis

Continuous variables of demographics and neuropsychological scores were presented as mean \pm standard deviation (SD) and compared by Student's *t* tests, or Mann-Whitney *U* test based on distributional properties. Categorical variables were described as percentages and compared by chi-square tests. Spearman's rank correlation test was performed to obtain the correlation coefficients between measured values or scores. The statistical tests were two-sided, and a P value < 0.05 was considered statistically significant. Analyses were conducted with SPSS version 25.0 (IBM Corp, Armonk, NY, USA).

Results

Demographic and Neuropsychological Tests

Clinical and demographic information for the subjects is shown in Table 1. No significant difference in gender, MMSE and MoCA scores was found between the MSA-c type and control groups. The patients exhibited significant higher UMSARS I and II scores.

Neuroimaging Analyses

Comparison of Cerebellar Atrophy in Patients Versus HCs by VBM Analysis

Atrophy in cerebellar subregions was apparent in patients relative to HCs. Specifically, one large and contiguous cluster was identified in the CBMc (voxel size: 10776; peak MNI coordinates: x = -7.5, y = -67.5, z = -34.5). Another large and contiguous cluster was identified in the CBMm (voxel size: 20158; peak MNI co-ordinates: x = -6, y = -67.5, z = -34.5). After FWE correction, comparisons between overall mean gray matter intensity of patients compared with those of HCs revealed an overall volume loss of 16.59% in the CBMc (P < 0.001) and 18.74% in the CBMm (P < 0.001) (Fig. 2).

 Table 1
 Demographic and clinical characteristics of the participants

Characteristics	MSA- C(<i>n</i> = 18)	Control(n = 18)	P value
Age, years	57.56±1.34	57.61±1.19	0.376*
Gender, male/female	8/10	7/11	0.735#
Education, years	13.78 ± 0.46	13.72 ± 0.49	0.183*
Disease duration, years	4.28 ± 0.19	NA	
MMSE	27.00 ± 1.68	27.56 ± 1.79	0.344*
MoCA	27.83 ± 1.20	28.28 ± 1.18	0.270*
UMSARS-I, total	18.39 ± 1.31	NA	
UMSARS-II, total	17.61 ± 1.41	NA	
Over disability grade	2.50 ± 0.22	NA	

MMSE Mini-Mental State Examination, *MoCA* Montreal Cognitive Assessment, *UMSARS* Unified Multiple System Atrophy Rating Scale, *NA* Not Applicable

*Student's t test or Mann-Whitney U test performed

Pearson's chi-square test performed

Resting-State Functional Connectivity Analysis

To investigate functional connectivity alterations in the patients, seed based interregional correlation analysis was performed. We extracted several large-scale cortical networks including aDMN, pDMN, IFPN,rFPN, dAN, and SMN using the ICA method. Then, we selected the two subregions of the bilateral cerebellum and analyzed the functional connectivity between the cerebellar subsystem and cortical networks.

After FWE correction, assessments for the patterns of resting-state cortico-cerebellar connectivity revealed disrupted connectivity between the CBMc and pDMN (patients: mean $r = 0.38 \pm 0.05$; HCs: mean $r = 0.63 \pm$ 0.07; t = -3.025; P = 0.005), rFPN (patients: mean r = 0.31 ± 0.07 ; HCs: mean r = 0.60 ± 0.05 ; t = -3.225; P = 0.003), and dAN(patients: mean $r = 0.41 \pm 0.07$; HCs: mean $r = 0.60 \pm 0.07$; t = -2.040; P = 0.005) in MSA-c patients. In addition, after FWE correction, decreased connectivity was also observed between the CBMm and cortical networks pDMN (patients: mean $r = 0.43 \pm$ 0.05; HCs: mean $r = 0.68 \pm 0.05$; t = -3.518; P = 0.001), IFPN (patients: mean $r = 0.67 \pm 0.06$; HCs: mean r = 1.03 ± 0.05 ; t = -4.580; P = 0.000), rFPN (patients: mean $r = 0.28 \pm 0.06$; HCs: mean $r = 0.45 \pm 0.05$; t = -2.087; P = 0.045, dAN (patients: mean $r = 0.44 \pm 0.07$; HCs: mean $r = 0.72 \pm 0.07$; t = -2.999; P = 0.005), and SMN (patients: mean $r = 0.52 \pm 0.06$; HCs: mean $r = 0.73 \pm$ 0.06; t = -2.399; P = 0.022) in the MSA-c patients (Fig. 3).

Within the CBMm and STN, patients displayed decreased connectivity when compared to HCs (patients: mean $r = 0.22 \pm 0.06$; HCs: mean $r = 0.42 \pm 0.07$; t = -2.151; P = 0.039, after correcting for FWE).

Relationship Between Cerebellar GM Intensity and the Resting-State Functional Connectivity

The extent of atrophy in the cerebellum was differentially correlated with specific impairments in resting-state connectivity between the cerebellar subsystem and the resting-state networks in patients, when compared with HCs.

Specifically, after Bonferroni-Holm correction, the extent of cerebellar atrophy was associated with a decrease in connectivity between the CBMc and the aDMN (r = -0.547, P = 0.023), pDMN (r = -0.496, P = 0.043), rFPN (r = -0.519, P = 0.033). With regard to the CBMm network, cerebellar atrophy was related to a significantly decreased connectivity between the CBMm and the aDMN (r = -0.637, P = 0.006), pDMN (r = -0.708, P = 0.001), rFPN (r = -0.558, P = 0.020), SMN (r = -0.506.9, P = 0.038), suggesting a potential relationship between atrophy and functional connectivity (please see Fig. 4).

After FWE correction, there was no significant difference between HCs and patients with atrophy in the STN (P > 0.500). At the same time, the degree of cerebellar atrophy was found to be related to the decrease in connectivity between the CBMm and the STN (r = -0.563, P = 0.019, after Bonferroni-Holm correction).

Relationship Between Motor Severity, Connectivity, and Cerebellar Atrophy

We observed an inverse correlation between the UMSARS II and the extent of cerebellar atrophy in the CBMm (r = -0.527, P = 0.030). No other relationships between atrophy of CBMc and connectivity were significantly correlated with UMSARS scores (please see Fig. 5).

Fig. 2 Voxel-based morphometry showing gray matter loss in the cognitive and motor cerebellar territories for MSA patients in comparison to controls. Areas of significant gray matter loss (red) in the cognitive and motor cerebellar territories, for patients with MSA versus control subjects. Results reported at P < 0.05, corrected for multiple comparisons (FWE)







Discussion

The present study aimed at seeking the relationship of structural and functional cerebellar GM abnormalities in MSA-c. Our findings revealed a significant decreased volume in several cerebellum subregions in MSA-c patients, and showed the extent of cerebellar atrophy was related to reduced functional connectivity between cerebellar subregions and largescale cortical networks including DMN, rFPN, dAN, IFPN, and SMN. Furthermore, we found abnormalities in motor performance to be associated with the extent of local cerebellar motor subsystem atrophy, which might be accompanied by decreased connectivity between the functional subsystems and large-scale cortical networks and STN. These results confirmed a great role of intrinsic structural and functional changes in the cerebellum in MSA-c and might improve our understanding of the neural pathophysiology mechanisms of MSAc from the perspective of network.

We observed significant atrophy in cerebellar motor and cognitive subregions in patients with MSA-c. Structurally, most MSA neuroimaging researches have focused on the gray



Fig. 4 The structure-function relationships from cerebellum to cortical network. Relationships between cerebellar atrophy and resting-state cerebellar functional connectivity. The solid lines denote a loss of connectivity. aDMN = anterior DMN; pDMN = posterior DMN; rFPN = right frontal parietal network; SMN = sensorimotornetwork

matter atrophy detected by the structural MRI [12, 14, 31–33]. Our study highlighted the atrophy of the cerebellar cognitive and motor subregions. The alpha synuclein-positive GCIs deposited in the cerebellum were found to destroy the neuron, which might lead to the atrophy of the cerebellum and cerebellar ataxia in patients with MSA-c. These findings might reflect the pathological changes of MSA- c [12, 14]. The novelty of this study was to explore the influence of cerebellar subregions loss on functional connectivity patterns within cerebellar-cortical networks in MSA-c. We did find a substantial impact of cerebellar atrophy on several decreased resting-state connectivity between cerebellar subsystems and large-scale cortical networks. The data on resting-state cerebello-



Fig. 5 Relationship between motor severity and atrophy.Relationship between cerebellar atrophy and clinical variables. We observed a negative correlation between the extent of atrophy as measured by section II of the UMSARS

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cortical functional connectivity are consistent with numerous researches showing its reduced-activation[15–17, 34, 35].Our results demonstrate that the cerebellar GM volume loss may be an important determinant of alterations in some cerebellocortical functional connectivity. Local volume reduction of cerebellar GM in the patients was related to a loss of connectivity between the motor cerebellum and aDMN, pDMN, and rFPN, as well as SMN, and between the cerebellar cognitive subsystem and aDMN and pDMN, as well as rFPN.

The rFPN is reported to play an important role in the executive control function, and the SMN was known involved in primary sensory and motor cortices[36]. We speculate that the disconnection between the motor cerebellum and these networks due to volume loss may be responsible for the control deficits and sensorimotor impairment of the disease. The DMN is known as a functional-anatomic network, which is involved in memory, self-reflection, and stream-ofconsciousness processing [37]. Interestingly, we observed the DMN involved in the hypoactive cortico-cerebellar motor circuitry in MSA-c. This result suggests the complicated mechanism of local cerebellar atrophy effecting motor deficits, in which many cortical networks associated with several functions including not only motor but also emotion [38], behavior[39, 40], and cognition[41, 42] is involved.

Cerebellar atrophy in MSA has been previously reported to correlate with behaviors [43]. Consistent with this result, our findings show a close relationship between motor severity and cerebellar motor subregion atrophy in MSA-c. As we know, the scores of UMSARS II can reflect the motor function. These correlations suggest that cerebellar motor subsystem atrophy may account for motor dysfunction in MSA-c due to pathological loss of connectivity with large-scale cortical networks. Since the correlations between motor cerebellum and cortical networks occur in networks with no direct anatomical connections, we investigate the association of motor cerebellar atrophy with the major input structure, the STN [44]. In our study, we find reduced-activation in the subthalamo-cerebellar functional connectivity without volume reduction in STN, suggesting that motor cerebellum change might influence connectivity between the cerebellar subgroups and large-scale cortical networks through the STN. This result highlighted that atrophy of the cerebellar motor subregion may be an initiating step in the hypoactive cortico-cerebellar motor circuitry.

In contrast to the bidirectional changes in connectivity between local cerebellar atrophy and cerebellar-cortical connectivity [45, 46], in the current study, no cortical networks were observed over recruited by patients, in a compensatory attempt. Our results provide evidence of no network-level compensatory mechanism in MSA-c, which consistent with other MSA neuroimaging researches[16, 35, 47]. Given the overactivation in cortical functional connectivity patterns induced by cerebellar atrophy in PD [46] and healthy aging [45], it is possible that massive degeneration may deprive a compensatory mechanism of MSA, whereas surviving regions may provide a compensatory mechanism [48]. In MSA-c, a wide range of neurons were influenced by the degeneration in the cerebellum, suggesting that no regions are left to provide a compensatory mechanism. Our findings emphasize the role of cerebellar atrophy in pathological pattern in MSA-c, which is different from those in PD or normal aging.

Even though the decreased connectivity patterns between the cognitive cerebellum and several cortical networks including DMN and rFPN were found in the present study, no cognitive performance differences were observed related to cerebellar atrophy between patients and HCs. These findings are inconsistent with previous reports, which showed the cognitive deficits in relation to the cerebello-cortical network contributions [34] or cerebellar volume reductions accounting for cognitive deficits^[43] in MSA. Unlike with pathology within the cerebral cortex in Alzheimer's disease, selective neuronal loss is found mainly present in the cerebellum and basal ganglia in MSA[49], thus cognitive impairment is not available for all MSA patients. Although the cerebellum plays a crucial role in cognitive regulation [50], the structural and functional abnormalities in the cognitive cerebellum contributing to cognitive manifestations were not found in the present study. These results allow us to take the hypothesis that the cerebral cortex or cortical networks play a major role in the neurobiological mechanisms of cognitive impairment in MSA-c, and volume loss in cognitive cerebellum which account for decreased cerebellocortical functional connectivity may be just a supplementary link in cognitive impairment. However, this hypothesis has yet to be verified.

Some of the limitations of our study are worth mentioning. First, in the current study, we mainly focused on the functional connectivity between cerebellar subregions and cortical networks. Further studies that simultaneously combine the resting-state and task-related fMRI, diffusion, perfusion, and metabolic data would be more helpful for a deep understanding of the MSA patients. Second, our study only focused on MSA-c. However, exploring MSA different subtype (parkinsonian and cerebellar variants) would provide valuable biomarkers for the early differential diagnosis of the disease. Third, MoCA and MMSE were administered as a measure of global cognitive performance without specific cognitive domains (eg. memory, mental speed, attention or executive function) being assessed, which might result in no significant difference in cognitive function being found between patients and HCs. Finally, a relatively small sample size was included in this study. Therefore, future longitudinal studies with a much larger sample are warranted to elucidate the progressive functional changes of the MSA patients and its relationship with the clinical performances.

Conclusion

Our findings reveal the relationship of atrophic and functional changes in the cerebellum in MSA-c. Furthermore, we have identified that cerebellar motor subsystem atrophy may account for motor deficit in MSA-c due to decreased connectivity with large-scale cortical networks and STN. These results not only provide compelling evidence that cerebellar atrophy is pivotal in neuropathological pathways of MSA-c, but also suggest the potential mechanism of cerebellar atrophy effecting motor dysfunction. These findings have important implications for the underlying neurobiology of MSA-c and added new evidence for the disconnection syndrome of MSA-c.

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Author Contributions Dr. Zhiqun Wang and Dr. Zhigang Chen had full access to all of the data in the study and takes responsibility for the integrity of the data and the accuracy of the data analysis.

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Compliance with Ethical Standards

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

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