Study of altered functional connectivity in individuals at risk for Alzheimer’s Disease

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Abstract.

BACKGROUND: Mild Cognitive Impairment (MCI) has been considered to have a high risk in converting into Alzheimer’s Disease (AD). Previous studies showed that AD was associated with changes in resting-state networks (RSNs). However, few studies have evaluated the altered functional connectivity in early mild cognitive impairment (EMCI) and late mild cognitive impairment (LMCI).

OBJECTIVE: The aim of this work was to evaluate the impaired network functional connectivity with the disease progression.

METHODS: In this paper, we evaluated the impaired function connectivity with the progression of disease based on a priori defined 246 regions of interest based on Brainnetome Atlas. Connectivity analysis based on three levels (node integrity, intra-network, and inter-network) was conducted.

RESULTS: Altered function connectivity was detected in several RSNs. These results provided insights into the dysfunction of more RSNs accompany the progression of AD. We also found that one brain region may belong to multiple RSNs and contribute to achieving different network function.

CONCLUSIONS: The aberrant intra- and inter-network dysfunctions might be potential biomarkers or predictors of MCI and AD progression and provide new insight into the pathophysiology of these diseases.

Keywords: Altered functional connectivity, early mild cognitive impairment, late mild cognitive impairment, functional connectivity

1. Introduction

Alzheimer’s Disease (AD) is a wide prevalent neurodegenerative disease, which endangers the health of human all over the world. The accumulations of A\textsubscript{β}1–42 and tau tangles have been considered as the biological markers of AD [1,2]. Clinically, more extensive and serious cognitive impairments were detected in the later stage than early stage of AD, and patients with mild cognitive impairment (MCI) have a large chance to change AD [3,4].

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With the development of neuroimaging techniques, MRI has been widely applied to investigate the neurodegenerative diseases. Resting-state fMRI (rsfMRI) is an especially useful modality because it can be used easily in diverse patient populations to quantify functional network connectivity in terms of correlated, spontaneous MR signal fluctuations [5]. It is efficient to assess brain function connectivity among brain regions [6,7]. Correlated brain regions have been divided into several resting-state networks (RSNs) [8,9].

The research about RSNs offered a typical model for looking into the relationship between brain activity alterations and cognitive dysfunction in neurological disorders [10]. As for as AD and MCI concerned, convergence evidence on account of fMRI studies demonstrated that functional integration have disease severity related changes. For examples, recent studies have shown that functional connectivity and networks were influenced in the subjects with AD or at high risk for AD, especially in higher cognitive RSNs such as the default mode network (DMN), attention network (AN), executive control network (ECN) and salience network (SAL). All these preliminary research findings support the point that AD is a disconnection syndrome [11] and that the selective impairment patterns in the RSNs may describe the traits and states of AD.

However, most of these studies concentrate on the connectivity alterations within higher cognitive RSNs, the understanding of more RSNs still needs to explore. Inspired by the above studies, we hypothesized that the alteration of functional connectivity across more RSNs would happen in individuals with EMCI and LMCI, which are patients at high risk for AD. We expected that functional abnormalities would occur in several RSNs and that the cognitive impairments measured by mini-mental state examination (MMSE) would be correlated with rs-fMRI markers of abnormalities. To test these hypotheses, in this paper, altered functional connectivities based on a priori defined 246 regions of interest [12] were explored in participants of 26 patients with EMCI, 19 patients with LMCI, compared with 23 healthy control subjects.

2. Materials and methods

2.1. Subjects

The dataset was examined from ADNI database. According to the protocol, each participant finished the informed consent and had neuropsychological tests and clinical examinations, including Clinical Dementia Rating (CDR), Mini-Mental State Examination (MMSE), and behavioral measures.

In this study, 68 participants with rs-fMRI tests were included. It includes 26 patients with EMCI and 19 patients with LMCI, compared with 23 healthy control subjects. The demographics of participants were showed in Table 1.

### Table 1: Demographics of subjects

|                | NC (23) | EMCI (26) | LMCI (19) | p value |
|----------------|---------|-----------|-----------|---------|
| Gender (M/F)   | 12/11   | 15/11     | 13/6      | 0.139   |
| CDR            | 0       | 0.5       | 0.5       | –       |
| Age (year)     | 74.0 (5.4) | 71.2 (7.9) | 73.2 (8.5) | 0.400   |
| MMSE           | 29.2 (1.0) | 28.3 (1.2) | 27.6 (1.2) | 0.002*  |
| Head motion    | 0.6 (0.4) | 0.7 (0.4) | 0.8 (0.6) | 0.800   |

*Indicates a statistical difference between groups, p < 0.05.
The inclusion criteria were as follows: 1) CDR of 0 and MMSE scores of 26 or higher for the NC group; 2) CDR of 0.5 and MMSE scores of 24 or higher for the MCI group; 3) no change in categorical disease status at the visit. The following exclusion criteria were applied: 1) obvious distortion on the raw imaging files; 2) translation or rotation in any axis of head motion larger than the size of one voxel during the scanning. For full ADNI inclusion/exclusion criteria, see http://www.adni-info.org.

2.2. Image acquisition

According to ADNI protocol (http://adni.loni.usc.edu/), the resting-state fMRI data was acquired on 3.0 Tesla Philips scanners (varied models/systems) while subjects were instructed to keep their eyes open. Images were acquired using an echo planar imaging (EPI) sequence with a repetition time (TR) ranging from 2250 ms to 3090 ms; echo time (TE) = 30 ms; flip angle (FA) = 80; slices = 48; field of view (FOV) = 212 mm RL, 198.75 mm AP, 159 mm FH. The imaging resolution in the X and Y dimensions ranged from 2.29 mm to 3.31 mm, and the slice thickness was 3.3 mm. For each subject, 140 volumes were acquired with total scan duration of 420 seconds. Some individuals received extended scans with 200 volumes for a total duration of 600 seconds. For consistency, only the first 140 volumes of these scans were analyzed.

2.3. Preprocessing of rs-fMRI

The resting-state fMRI data were pre-processed using the Brainnetome fMRI Toolkit (http://brant.brainnetome.org/) based on Statistical Parametric Mapping (SPM12, http://www.fil.ion.ucl.ac.uk/spm). Pre-processing steps included (1) slice-timing correction; (2) realignment to the first volume for head motion correction; (3) spatial normalization to a standard EPI template with re-slicing to 2 mm cubic voxels; (4) regression of nuisance parameters including linear drift, six motion parameters, and the mean time series of all voxels within the white matter and cerebrospinal fluid; (5) temporal filtering (0.01–0.08 Hz) to reduce high frequency and low frequency noise.

2.4. Connectivity analysis

The seed region in the present study were based on a priori defined 246 regions of interest (Brainnetome Atlas, http://atlas.brainnetome.org/) [12]. These regions of interests have been coregistered to the templates from Stanford’s Functional Imaging in Neuropsychiatric Disorders (FIND) (http://findlab.stanford.edu/functional_ROIs.html), which include Auditory Network (AN); Basal Ganglia Network (BGN); Dorsal Default Mode Network (dDMN); Higher Visual Network (HVN); Language Network (LN); Left Executive Central Network (LECN); Sensorimotor Network (SN); Posterior Salience Network (PSN); Precuneus Network (PN); Primary Visual Network (PVN); Right Executive Control Network (RECN); Anterior Salience Network (ASN); Ventral Default Mode Network (VDMN); Visuospatial Network (VN).

For each participant, Pearson’s correlation coefficient was computed between the mean time series of each pair of ROIs. Three levels investigation was explored [13]:

1. Node level: \( Z_{\text{integrity}} = \sum_{i=1}^{246} |Z_{ij}| \), where \( Z_{ij} \) stands for the score of Z between the \( i^{\text{th}} \) and \( j^{\text{th}} \) ROI.

2. Inter-network level: \( Z_X = \frac{1}{\text{Num}_X(\text{Num}_X-1)^{1/2}} \sum_{i,j \in X} |Z_{ij}| \), where X stands for a RSN, and Num\(_X\) stands for the numbers of ROIs in X, \( i, j \) stands for two different ROIs.
Fig. 1. The altered region between EMCI and LMCI at node level.

(3) Intra-network level: \( Z_{X,Y} = \frac{1}{\text{Num}_x \times \text{Num}_y} \sum_{i \in X, j \in Y} |Z_{i,j}| \), where X, Y stands for different networks of the 14 RSNs.

2.5. Statistical analysis

In order to obtain the statistical significance, the Z scores at each level were employed into one-way analyses of variance (ANOVA) with a group as a factor (3 groups: NC, EMCI, LMCI) after age and gender effects were regressed out by using a linear regression model. The statistical significance was \( P < 0.05 \).

In three groups, considering that we implemented many times comparisons, we performed a 10,000 times random arrangement to test whether the identified altered connections are really significant. For the sake of getting a statistically significant difference, post-hoc in each two groups have been performed between the NC and EMCI, NC and LMCI, EMCI and LMCI (\( P < 0.05 \)). To explore the relationship between functional connectivity and cognitive ability, we also computed the Pearson’s correlations between the Z scores and the MMSE scores at each level in the EMCI, and LMCI group.

3. Results

For each group, a 246 × 246 functional connectivity matrix was computed.

At node level, significant differences between groups were detected, and these differences were widely separated in several RSNs. (1) Compared with EMCI group and LMCI group (Fig. 1), significant decrease integrity was showed in two regions of Posterior Salience Network (PCL_R_2_1, INS_R_6_1), and two regions of Language Network (pSTS_L_2_2, MTG_L_4_3), and two regions of dorsal Default Mode Network (CG_L_7_1, CG_L_7_6). (2) Compared with EMCI group and NC group (Fig. 2), significant decrease integrity was showed in the region of Dorsal Default Mode Network (Hipp_L_2_1). (3) Compared with LMCI group and NC group (Fig. 3), significant decrease integrity was showed in five regions of dorsal Default Mode Network (CG_L_7_1, CG_L_7_6, SFG_L_7_7, IPL_L_6_5, Hipp_L_2_1), and one region of Anterior Salience Network (IFG_R_6_3), three regions of Language Network (IFG_L_6_3, pSTS_L_2_2, MTG_L_4_3), one region of Anterior Salience Network (IFG_R_6_3), three regions of Language Network (IFG_L_6_3, pSTS_L_2_2, MTG_L_4_3), one region of Basal Ganglia Network (BG_L_6_6), one region of Left Executive Control Network (MFG_L_7_2), one region of Sensorimotor Network (PrG_L_6_4).

At intra-network level (Fig. 4), compared with EMCI group, significant decreases were found in Auditory Network (AN) and Sensorimotor Network (SN) in the LMCI group. Compared with NC group,
Fig. 2. The altered region between EMCI and NC at node level.

the LMCI group showed significant decreases in Auditory Network (AN), Language Network (LN), Sensorimotor Network (SN).

At inter-network level (Fig. 5), compared with EMCI group, the connectivity strength difference between Auditory Network and Language Network (AN-LN) was found in LMCI. Compared with NC group, several connectivity strength differences were detected in LMCI group, such as Language Network and Left Executive Control Network (LN-LECN), Dorsal Default Mode Network and Language Network (dDMN-LN), Language Network and Anterior Salience Network (LN-ASN); Auditory Network-Language Network (AN-LN).

4. Discussion

In the present study, impaired functional connectivity was detected widely in MCI which might be the basis of cognitive impairment. At node level, the differences were anchored in several networks (Figs 1–3), such as Posterior Salience Network, Language Network, dorsal Default Mode Network, Anterior Salience Network Basal Ganglia Network Left Executive Control Network, Sensorimotor Network.

At intra-network level, the significant decreases were anchored in Auditory Network (AN), Language Network (LN), Sensorimotor Network (SN) in MCI (Fig. 4). At inter-network level, the significant differences were anchored in between Auditory Network (AN) and Language Network (LN), between Language Network (LN) and Left Executive Control Network (LECN); between Dorsal Default Mode Network (dDMN) and Language Network (LN); between Language Network (LN) and Anterior Salience Network (ASN) in MCI (Fig. 5).

In this study, the altered function connectivity was detected in AN and SN, both of which were parts of SMN (sensory-motor network) [14]. Previous studies have found that the functional changes among olfaction, visual, hearing and motor systems (the major components of the SMN) might influence the onset of cognitive impairments, worsen as the disease progress, and be strong risk factors for AD [15]. The language network was anchored in Wernicke’s and Broca’s areas which was responsible for language comprehension and production. Our results provided additional convincing evidence that the dysfunction in SMN should be paid more attention.

In inter-network level, changes were aroused in between LN and LECN, as well as dDMN and LN. The Executive Control Network (ECN) was involved in goal-directed behavior and cognitive control. The
Fig. 3. The altered region between LMCI and NC at node level.
DMN played a pivotal role in cognitive processes, especially in episodic memory processing [16,17]. The impaired performance of episodic memory was one of the core features in the diagnosis of AD [18]. And the impaired functional connectivity within DMN was frequently identified in AD and MCI patients using multiple imaging techniques [18–20]. Our results also demonstrated that DMN might be one of the most affected networks in MCI patients.

Previous resting state fMRI studies demonstrated aberrant functional connectivity of DMN and SAL.
in AD [20,21], aMCI [22] and even in normal cognition with amyloid burden [23]. It was suggested that the SAL was involved in the orientation of attention to the most homeostatically relevant (salient) of ongoing intrapersonal and extra personal events [24]. In this light, the present study prompted us to infer that the reduced connectivity between DMN and SAL might be one of the reasons for attention deficits in the early stages of AD and MCI.

These results provided insights into the dysfunction of more RSNs accompany the severity of subjects at high risk for AD. The aberrant intra- and inter-network dysfunctions might be potential biomarkers or predictors of MCI and AD progression.

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Conflict of interest

None to report.

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