Cerebral atrophy in amyotrophic lateral sclerosis parallels the pathological distribution of TDP43

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Amyotrophic lateral sclerosis is a neurodegenerative disease characterized by a preferential involvement of both upper and lower motor neurons. Evidence from neuroimaging and post-mortem studies confirms additional involvement of brain regions extending beyond the motor cortex. The aim of this study was to assess the extent of cerebral disease in amyotrophic lateral sclerosis cross-sectionally and longitudinally and to compare the findings with a recently proposed disease-staging model of amyotrophic lateral sclerosis pathology. Deformation-based morphometry was used to identify the patterns of brain atrophy associated with amyotrophic lateral sclerosis and to assess their relationship with clinical symptoms. Longitudinal T1-weighted MRI data and clinical measures were acquired at baseline, 4 months and 8 months, from 66 patients and 43 age-matched controls who participated in the Canadian Amyotrophic Lateral Sclerosis Neuroimaging Consortium study. Whole brain voxel-wise mixed-effects modelling analysis showed extensive atrophy patterns differentiating patients from the normal controls. Cerebral atrophy was present in the motor cortex and corticospinal tract, involving both grey matter and white matter, and to a lesser extent in non-motor regions. More specifically, the results showed significant bilateral atrophy in the motor cortex and corticospinal tract (including the internal capsule and brainstem) and ventricular enlargement, along with significant longitudinal atrophy in precentral gyrus, frontal and parietal white matter, accompanied by ventricular and sulcal enlargement. Atrophy in the precentral gyrus was significantly associated with greater disability as quantified with the Amyotrophic Lateral Sclerosis Functional Rating Scale-Revised ($P < 0.0001$). The pattern of atrophy observed using deformation-based morphometry was consistent with the Brettschneider’s four-stage pathological model of the disease. Deformation-based morphometry provides a sensitive indicator of atrophy in Amyotrophic lateral sclerosis and has potential as a biomarker of disease burden, in both grey matter and white matter.

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Abbreviations: ALS = amyotrophic lateral sclerosis; ALSFRS-R = ALS Functional Rating Scale-Revised; DBM = deformation-based morphometry; ECAS = Edinburgh Cognitive and Behavioral ALS Screen; FDR = false discovery rate

Graphical Abstract

Introduction

Amyotrophic lateral sclerosis (ALS) is a heterogeneous and progressive disorder, characterized by the degeneration of both the upper motor neurons arising from the cortex and lower motor neurons in the brainstem and spinal cord. There is, additionally, extensive involvement of extra-motor regions, including the frontal and temporal lobes (Foerster et al., 2013; Hardiman et al., 2017; Agosta et al., 2018).

The onset of ALS symptoms is variable and can include involvement of limb, bulbar or respiratory muscles or cognitive impairment. Spasticity and hyperreflexia result from upper motor neuron degeneration (Gordon, 2013), whereas fasciculations, cramps and muscle wasting result from lower motor neuron degeneration (Hardiman et al., 2017). Cognitive dysfunction also affects more than half of the patients, with evident frontotemporal dementia in 13% (Wilson et al., 2001; Ringholz et al., 2005). Approximately 30% of all patients have evidence of executive dysfunction at onset, and frontotemporal dementia is one of the presenting features in 13% of incident cases (Hardiman et al., 2017).

Structural MRI studies have demonstrated cross-sectional brain changes in the grey matter and white matter arising from ALS, in both motor and extra-motor areas (Kato et al., 1993; Oba et al., 1993; Waragai, 1997; Andreadou et al., 1998; Hecht et al., 2001; Charil et al., 2009; Foerster et al., 2013; Agosta et al., 2018). Voxel-based morphometry studies report grey matter loss in the motor cortex, as well as in frontotemporal regions especially for patients with cognitive deficits (Chang et al., 2005; Thivard et al., 2007; Agosta et al., 2009a, 2007; Senda et al., 2011; Foerster et al., 2013; Sheng et al., 2015; Shen et al., 2016; Buhour et al., 2017; Kim et al., 2017; Agosta et al., 2018; Menke et al., 2018; Chen et al., 2018b). Cortical thickness studies have reported cortical thinning in the motor cortex as well as frontal and temporal areas (Roccatagliata et al., 2009; Verstraete et al., 2012; Chen et al., 2018c). These studies have generally been performed in relatively small samples of patients and, in particular, longitudinal studies are sparse (Agosta et al., 2009a, b; Menke et al., 2014, 2018), inherent in this rare and rapidly progressive condition, resulting in challenges with patient recruitment and retention (Foerster et al., 2013; Agosta et al., 2018). Furthermore, insufficient statistical power and differences in image-processing pipelines and statistical analyses have led to variabilities in the reported findings, with studies reporting diverging results such as focal atrophy in
motor/premotor regions, widespread frontotemporal grey matter atrophy sparing the motor cortex or no significant atrophy (Foerster et al., 2013; Agosta et al., 2018).

In this study, we have used deformation-based morphometry (DBM) to quantify the patterns of disease-related brain changes in ALS. DBM allows simultaneous characterization of both mesoscopic and macroscopic changes in brain anatomy in grey matter, white matter and cerebrospinal fluid (Ashburner et al., 1998). In addition, the image-processing tools used in this study have been designed for use in multi-centre datasets across different MRI systems and are able to accommodate between-site variabilities. Taking advantage of these robust methodological tools, our primary goal was to accurately quantify the patterns of disease-related brain changes in ALS and investigate the associations between such changes and clinical symptoms.

In addition, these in vivo changes were investigated in relation to the pathological staging system of Brettschneider et al. (2013) that characterizes ALS in four stages based on the regional accumulation of transactive response DNA-binding protein 43 kDa. Stage 1 is characterized by the involvement of the primary motor cortex, alpha motor neurons in the ventral horn of the spinal cord and brainstem motor nuclei of cranial nerves V, VII and X–XII. Stage 2 adds involvement of the prefrontal neocortex, reticular formation and the inferior olivary complex. In Stage 3, transactive response DNA-binding protein 43 kDa deposition is now present more extensively in the prefrontal and postcentral neocortex and striatum. Stage 4 additionally involves the anteromedial portions of the temporal lobe, including the hippocampal formation.

We used longitudinal MRI and clinical data from 66 ALS patients and 43 age-matched controls who participated in the Canadian ALS Neuroimaging Consortium to investigate the pattern of brain atrophy in ALS. Three specific hypotheses are addressed: (i) cerebral atrophy is widespread in ALS, involving grey matter, white matter and motor and extra-motor regions; (ii) regional grey and white matter atrophy is correlated with motor and cognitive symptoms; and (iii) the spatial pattern of in vivo atrophy aligns with the Brettschneider stages of ALS pathology.

Materials and methods

Participants

Data used in this study included longitudinal (baseline, Month 4 and Month 8) MRI and clinical measurements of 66 ALS patients ($N_{Baseline} = 64$, $N_{Month4} = 44$, $N_{Month8} = 24$) as well as 43 age-matched healthy controls ($N_{Baseline} = 42$, $N_{Month4} = 32$, $N_{Month8} = 21$) from the Canadian ALS Neuroimaging Consortium (http://calsnic.org/, ClinicalTrials.gov NCT02405182). Each participating centre in Canadian ALS Neuroimaging Consortium followed identical standard operating procedures for clinical evaluations and harmonized acquisition protocols for brain imaging. Data from four sites were included: University of Alberta, University of Calgary, University of Toronto and McGill University. All participants gave written informed consent, and the study was approved by the health research ethics boards at each of the participating sites. Participants were included if they had sporadic or familial ALS and met El Escorial criteria for possible, probable, probable laboratory-supported or definite ALS (Brooks et al., 2000). Participants were excluded if they had a history of other neurological or psychiatric disorders, prior brain injury or respiratory impairment resulting in an inability to tolerate the MRI protocol.

Clinical evaluations included global measures of disease status, upper motor neuron dysfunction and cognitive batteries. Disability was assessed with the ALS Functional Rating Scale-Revised (ALSFRS-R), and the annualized disease progression rate was estimated using the formula: $(48 - \text{ALSFRS-R})/\text{symptom duration}$. Finger and foot tapping rates (left and right) were measured and averaged as indicators of upper motor neuron function. Cognitive function was assessed using the Edinburgh Cognitive and Behavioral ALS Screen (ECAS), a multi-domain cognitive screening battery developed for patients with ALS (Abrahams et al., 2014).

The ECAS total score was used to dichotomize patients to impaired versus normal general cognitive status using a cut-off threshold of 2 SDs from the mean of the control subjects. Patients were also dichotomized to impaired versus normal on executive function (ALSexi) based on relevant sub-scores of the ECAS: subjects were considered executively impaired if either the ECAS verbal fluency or ECAS executive score was 2 SDs below the mean of the respective scores in control subjects. Note that the ECAS verbal fluency score is a composite score from two letter fluency tasks and the ECAS executive score is a composite score from tasks of reverse digit span, alternation, sentence completion and social cognition.

All imaging data were acquired on 3 T MRI systems. University of Alberta and McGill University acquired data using Prisma and TimTrio Siemens systems, respectively, and University of Toronto and University of Calgary acquired data using General Electric Healthcare (Discovery MR750) systems. The MRI acquisition protocol included three-dimensional T1-weighted scans acquired at 1 mm$^3$ isotropic resolution. Acquisition parameters were identical for the Siemens systems, which used a magnetization prepared rapid gradient echo sequence with repetition time $= 2300$ ms, echo time $= 3.43$ ms, inversion time $= 900$ ms, flip angle $= 9^\circ$, field of view $= 256 \text{mm} \times 256 \text{mm}$. The two General Electric systems used identical parameters as well for an inversion recovery-prepared fast spoiled gradient-recalled echo imaging sequence with repetition time $= 7.4$ ms, echo
time = 3.1 ms, inversion time = 400 ms, flip angle = 11°, field of view = 256 mm × 256 mm.

**MRI processing**

All T1-weighted MRI data were pre-processed using the Medical Imaging Network Common dataform toolkit of the Montreal Neurological Institute, publicly available at https://github.com/BIC-MNI/minc-tools, using the following steps: (i) denoising (Coupe et al., 2008); (ii) intensity inhomogeneity correction (Sled et al., 1998); and (iii) image intensity normalization into intensity range (0–100) according to a linear histogram matching algorithm. All images were first linearly (Dadar et al., 2018a) and then non-linearly (Avants et al., 2008) registered to an average template recognized by the International Consortium for Brain Mapping, Montreal Neurological Institute-International Consortium for Brain Mapping 152 (Fonov et al., 2009). The quality of the registrations was visually assessed, confirming that all imaging data were accurately registered to the template.

Using Medical Imaging Network Common tools, DBM maps were obtained by computing the Jacobian of the estimated non-linear deformation field. DBM maps reflect the local differences between the average Montreal Neurological Institute-International Consortium for Brain Mapping 152 template and images of a given participant, with values larger than one indicating expansion (e.g. larger sulci or ventricles) and values smaller than one indicating shrinkage (e.g. atrophy). Voxel-wise DBM maps were used to assess the global brain differences between controls and ALS patients. The mean DBM values across regions of interest were used to assess the relationship between regional atrophy and clinical measurements.

**Statistical analyses**

To investigate the disease-related cross-sectional and longitudinal changes, normative aging as well as potential sex differences were regressed out from the voxel-wise ALS patient data using the voxel-wise DBM data from the matched controls, as similarly performed in previous studies (La Joie et al., 2012; Moradi et al., 2015; Zeighami et al., 2019; Brown et al., 2019). To achieve this, the following voxel-wise mixed-effects models were estimated based on the controls:

\[
DBM \sim 1 + \text{age} + \text{sex} + (1|\text{ID}) + (1|\text{centre}),
\]

(1) where DBM indicates the DBM value at a certain voxel in the brain; age indicates the participant’s age at the time of imaging; sex (male/female) is a categorical variable; and ID and centre are categorical random variables, indicating the control participant’s ID and the data acquisition centre, respectively. Based on the model estimates, normative aging and sex effects were then regressed out from the patients, yielding residualized DBM values (also known as W-score maps) representing the deviation of the patient measures relative to the value expected in the control group for patient’s age and sex (La Joie et al., 2012; Brown et al., 2019). The following mixed-effects models were used to estimate the ALS-specific brain changes in the patients, accounting for confounds:

\[
DBMr \sim 1 + \text{age} + \text{sex} + (1|\text{ID}) + (1|\text{centre}).
\]

(2)

Here, DBMr are the residualized DBM values [a.k.a. W-score maps, obtained after regressing out age and sex based on controls, i.e. (1)]. The resulting maps were corrected for multiple comparisons using the Benjamini and Hochberg/Yekutieli false discovery rate (FDR) controlling method (Benjamini and Hochberg, 1995; Benjamini and Yekutieli, 2001) with a significance threshold of 0.05.

In addition, the mean DBM values were calculated for the corticospinal tract mask obtained from Yeh et al. (2018), as well as the frontal lobe, frontotemporal lobe, precentral gyrus and dorsolateral prefrontal cortex masks obtained from a grey matter atlas for the Montreal Neurological Institute-International Consortium for Brain Mapping 152 template (Manera et al., 2019). The average values were used in the following region-of-interest analyses:

\[
DBMr \sim 1 + \text{clinical measure} + (1|\text{ID}) + (1|\text{centre}),
\]

(3) where clinical measure indicates either categorical dichotomized scores [e.g. median progression rate (fast/slow)] or continuous values [e.g. ALSFRS-R (0–48)]. The decision on dichotomization of general cognition (ECAS) and executive function (ALSexi) was made due to the fact that the ECAS was performed only at baseline, and therefore, we were not able to measure longitudinal change in these variables. Table 1 indicates the list of the regions and measures of interest tested. The results were corrected for multiple comparisons using the Benjamini and Hochberg/Yekutieli FDR controlling method with a significance threshold of 0.05 (Benjamini and Hochberg, 1995; Benjamini and Yekutieli, 2001). All analyses were performed using MATLAB software (version 2019b). The analyses were performed using ‘fitlm’ tool, appropriate for performing linear mixed-effects modelling of longitudinal data and accounting for the random effects.

**Disease staging based on atrophy**

The t-statistics W-score map was incrementally thresholded at higher levels to reveal increasing atrophy, starting at −2 (FDR corrected significance level) with one-point increments (i.e. −3, −4 and −5, equivalent to half of the standard deviation from the mean of the t-statistics distribution). Four stages with increasing atrophy were identified. The results were then visually compared with the ALS stages defined based on previous pathological staging studies (Brett Schneider et al., 2013), based on the regions involved at each stage.
Data availability statement
Anonymized data will be shared at the request of qualified investigators.

Results
Table 2 summarizes the demographic information as well as the clinical scores for the participants used in this study.

Figure 1 shows the $t$-statistics maps reflecting the significant brain volume changes in the ALS cohort, i.e. deviations of the patient DBM measures relative to the value expected in the control group for patient's age and sex, derived from the intercept term in Model 2, after FDR correction. Warm colours indicate areas that are significantly enlarged in the patients in comparison with the controls (e.g. in the sulci and ventricular regions), and cold colours indicate areas of additional shrinkage (i.e. atrophy) in the patients. The results show significant atrophy in the precentral gyrus that is more pronounced on the right, as well as diffuse atrophy in frontal and parietal white matter and ventricular and sulcal enlargement.

Table 3 shows the associations between clinical measures and regional atrophy. The mean DBM values in the precentral gyrus were significantly and positively associated with ALSFRS-R, indicating that a lower DBM value (i.e. greater atrophy) related to a lower score in ALSFRS-R (i.e. greater disability). Similarly, the mean DBM value in the corticospinal tract was positively associated with ALSFRS-R, but this relationship did not survive multiple comparison correction (uncorrected $P$-value $= 0.03$). There was also a trend towards an association between lower mean DBM value in the dorsolateral prefrontal cortex and greater executive impairment (uncorrected $P$-value $= 0.08$).

Disease staging based on atrophy
$t$-Statistic thresholds of the W-score map (Fig. 1) at $-2$, $-3$, $-4$ and $-5$ (FDR corrected and each equivalent to half of the standard deviation from the mean $t$-statistic distribution) revealed increasing magnitude and spatial extent of atrophy that closely corresponded with the Brettschneider et al. (2013) pathological stages (Fig. 3).
Figure 1 ALS atrophy pattern. Sagittal, coronal and axial slices showing the t-statistic maps reflecting the significant patterns of brain volume changes in the ALS cohort [corrected for expected age and sex changes using (1)]. Warm colours indicate enlargement (e.g. ventricular and sulcal regions), and cold colours indicate shrinkage of the tissue (i.e. atrophy). X, Y and Z values indicate MNI coordinates for the displayed slice.

Figure 2 Longitudinal changes. Sagittal, coronal and axial slices showing the t-statistic maps reflecting the significant brain volume changes with age (i.e. over time) in the ALS cohort. Warm colours indicate enlargement (e.g. ventricular and sulci regions), and cold colours indicate shrinkage of the tissue (i.e. atrophy). X, Y and Z values indicate MNI coordinates for the displayed slice.
Discussion

The objective of this study was to characterize longitudinal cerebral atrophy in ALS in vivo. In line with our hypotheses, it was revealed that atrophy in ALS is extensive, involving bilaterally the motor cortex and corticospinal tracts including the internal capsule and brainstem, and is accompanied by an overall pattern of ventricular enlargement. Longitudinally, atrophy progressed in bilateral precentral gyri, diffuse white matter, and was accompanied by ongoing ventricular and cortical sulcal enlargement. Furthermore, atrophy in the precentral gyrus significantly correlated with disability as assessed with the ALSFRS-R. Finally, the pattern of atrophy is consistent with the pathologically defined stages in ALS by Brettschneider et al. (2013). To achieve these results, DBM was applied to T1-weighted scans from a large multi-centre cohort of ALS from Canadian ALS Neuroimaging Consortium.

Previous imaging studies have reported similar patterns of atrophy. For example, voxel-based morphometry studies have reported gray matter (GM) reduction in motor, premotor, basal ganglia and frontotemporal regions (Chang et al., 2005; Agosta et al., 2007; Thivard et al., 2007; Agosta et al., 2009a; Senda et al., 2011; Shen et al., 2015; Shen et al., 2016; Buhour et al., 2017; Kim et al., 2017; Chen et al., 2018b; Menke et al., 2018). A longitudinal tensor-based morphometry study of 16 ALS patients and 10 matched controls (9 months follow-up) found progression of GM atrophy in the left premotor cortex and right basal ganglia in the patients as well as accelerated GM loss in motor and prefrontal areas in patients with faster clinical progression (Agosta et al., 2009a). Cortical thickness studies have reported cortical thinning in the motor cortex as well as frontal and temporal areas (Roccatagliata et al., 2009; Verstraete et al., 2012; Chen et al., 2018c). A recent meta-analysis

Table 3 Summary of the results of the mixed-effects models of associations between regional DBM values and clinical measures

| Region                      | t-Statistics | P-value |
|-----------------------------|--------------|---------|
| UMN burden                  |              |         |
| Corticospinal tract         | -0.03        | 0.97    |
| Precentral gyrus            | -0.33        | 0.74    |
| Progression rate            |              |         |
| Corticospinal tract         | 1.18         | 0.24    |
| Precentral gyrus            | 1.29         | 0.20    |
| General cognitive status (ECAS) |   |         |
| Frontotemporal lobe         | 1.39         | 0.17    |
| ALSexi                      |              |         |
| Dorsolateral prefrontal cortex | -1.72   | 0.08    |
| Site of onset               |              |         |
| Corticospinal tract         | -0.56        | 0.58    |
| Precentral gyrus            | -1.11        | 0.27    |
| Tapping rate                |              |         |
| Corticospinal tract (contralateral) | 0.73 | 0.46    |
| Precentral gyrus            | 1.30         | 0.19    |
| ALSFRS-R                    |              |         |
| Corticospinal tract         | 2.08         | 0.03    |
| Precentral gyrus            | 4.13         | <0.0001 |
| Frontotemporal lobe         | 0.80         | 0.42    |

Significant results after correction for multiple comparisons are displayed in bold font. ALSexi = executive impairment; UMN = upper motor neuron.

Figure 3 Disease staging. Comparison of disease stages based on the magnitude of atrophy and histological stages. The first row indicates the four pathological stages based on Brettschneider et al. (2013) as well as a corresponding sagittal slice from the t-statistics map (Fig. 2). The second row shows three sagittal, coronal and axial slices overlayed with the t-statistics map values thresholded at t-values = -5, -4, -3 and -2, to correspond with Stages 1–4, respectively.
reported significant WM reduction in the bilateral supplementary motor areas, precentral gyri, left middle cerebral peduncle and right cerebellum, involving the corticospinal tract, interhemispheric fibres, subcortical arcuate fibres and the projection fibres to the striatum and corticoponto-cerebral tract (Chen et al., 2018a).

The atrophy pattern observed in this study provides an in vivo confirmation for the sequentially increasing pathology in alignment with Brettschneider model (Brettschneider et al., 2013). This sequential pattern of disease progression has also been reported in diffusion tensor imaging studies (Kassubek et al., 2014; Müller et al., 2016; Schmidt et al., 2016). Further studies are necessary to determine the nature of the underlying biological changes (microstructural damage, macrostructural atrophy or a combination) that are observed in T1-weighted and diffusion tensor imaging (DTI) data. To our knowledge, no previous morphometric study of ALS has succeeded to show the correspondence between the pattern of grey and white matter atrophy and disease stages defined by transactive response DNA-binding protein 43 kDa spreading (Brettschneider et al., 2013). Using different threshold values on the t-statistics map (reflecting the severity of atrophy specific to ALS), the four stages described by Brettschneider et al. were identified (Fig. 3). The possibility to determine disease stage, and pathological propagation, using structural MRI could have potential implications in the clinical setting. Indeed, using already available and routinely used MRI, non-motor progression of symptoms (i.e. cognitive impairment) could be anticipated. Further studies are necessary to investigate the sensitivity of using structural MRI at the level of an individual participant.

The present study has several advantages compared with the previous reports. The multi-centre Canadian ALS Neuroimaging Consortium cohort included 66 ALS patients, a larger sample size compared with previous studies (Sheng et al., 2015). The MRI scans were also acquired with higher magnetic field strength (i.e. 3T rather than 1.5T), yielding images with better contrast. All the image-processing tools used in this study have been developed and extensively validated for use in multi-centre studies involving different MRI systems and have been used in numerous such studies (Zeighami et al., 2015; Dadar et al., 2018b; Misquitta et al., 2018; Dadar et al., 2019; Manera et al., 2019; Sanford et al., 2019). In addition, all analyses included ‘centre’ as a categorical random effect to ensure that any residual variability across centres would not bias the results.

We acknowledge that there were limitations to the present study. Due to the challenges in enrolment and follow-up of the patients in this rare and rapidly progressing disease, the sample sizes of ALS studies (including ours) are in general small compared with other more common neurodegenerative diseases. Attrition is also another common challenge in ALS studies. Not all participants completed all three visits, although the proportion of the number of follow-up visits to baseline was similar across patient and control groups. Furthermore, our follow-up duration was relatively short (8 months). However, since patients with severe symptoms and faster disease progression tend to not complete the follow-up visits, making studies with long follow-up durations biased towards the healthier patients with slower progression rates (Foerster et al., 2013), this relatively short follow-up might have had the advantage of making our findings less biased towards the healthier patients who are more likely to complete longer follow-ups, leading to more generalizable results. Future studies investigating DBM changes in ALS with longer follow-ups are warranted, though there are inherent constraints in recruiting such a cohort.

In conclusion, DBM reveals the atrophy pattern characteristic of ALS consistent with previous pathological findings, both in grey and white matter areas, and can be used to obtain a quantitative measure of disease burden. DBM measurements might therefore have the potential to be a biomarker in ALS.

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Competing interests
The authors report no competing interests.

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