Title
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Permalink
https://escholarship.org/uc/item/2xj3q6wg

Journal
Neurology(R) neuroimmunology & neuroinflammation, 6(6)

ISSN
2332-7812

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Publication Date
2019-11-01

DOI
10.1212/nxi.0000000000000616

Peer reviewed
Contribution of normal aging to brain atrophy in MS

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Neurol Neuroimmunol Neuroinflamm 2019;6:e616. doi:10.1212/NXI.0000000000000616

Abstract

Objective
To identify the top brain regions affected by MS-specific atrophy (i.e., atrophy in excess of normal aging) and to test whether normal aging and MS-specific atrophy increase or decrease in these regions with age.

Methods
Six hundred fifty subjects (2,790 MRI time points) were analyzed: 520 subjects with relapse-onset MS from a 5-year prospective cohort with annual standardized 1-mm 3D T1-weighted images (3DT1s; 2,483 MRIs) and 130 healthy controls with longitudinal 3DT1s (307 MRIs). Rates of change in all FreeSurfer regions (v5.3) and Structural Image Evaluation Using Normalization of Atrophy (SIENA) were estimated with mixed-effects models. All FreeSurfer regions were ranked by the MS-specific atrophy slope/standard error ratio ($\beta_{\text{MS}} \times \text{time}/\text{SE}_{\beta_{\text{MS}} \times \text{time}}$). In the top regions, age was added as an effect modifier to test whether MS-specific atrophy varied by age.

Results
The top-ranked regions were all gray matter structures. For SIENA, normal aging increased from 0.01%/y at age 30 years to −0.31%/y at age 60 years (−0.11% ± 0.032%/decade, $p < 0.01$), whereas MS-specific atrophy decreased from −0.38%/y at age 30 years to −0.12%/y at age 60 years (0.09% ± 0.035%/decade, $p = 0.01$). Similarly, in the thalamus, normal aging increased from −0.15%/y at age 30 years to −0.62%/y at age 60 years (−0.16% ± 0.079%/decade, $p < 0.05$), and MS-specific atrophy decreased from −0.59%/y at age 30 years to −0.05%/y at age 60 years (0.18% ± 0.08%/decade, $p < 0.05$). In the putamen and caudate, normal aging and MS-specific atrophy did not vary by age.

Conclusions
For SIENA and thalamic atrophy, the contribution of normal aging increases with age, but does not change in the putamen and caudate. This may have substantial implications to understand the biology of brain atrophy in MS.
Glossary

ADNI = Alzheimer’s Disease Neuroimaging Initiative database; CIS = clinically isolated syndrome; DMT = disease-modifying therapy; DTI = diffusion tensor imaging; HC = healthy control; PBVC = percentage whole-brain volume change; RRMS = relapsing-remitting MS; SIENA = Structural Image Evaluation Using Normalization of Atrophy; SPMS = secondary progressive MS; UCSF = University of California San Francisco.

The field of MS has devoted substantial effort to understanding and quantifying whole-brain and regional gray matter atrophy using MRI. Although the exact mechanism is unclear, brain atrophy in MS is thought to represent the net accumulation of tissue damage over time. Using MRI, numerous cross-sectional and longitudinal studies have demonstrated the clinical relevance of whole- and regional brain volume loss in MS. As such, brain volumetrics, particularly whole-brain volume, are often incorporated into MS clinical trials and clinical research as an in vivo proxy of neurodegeneration.

Historically, the literature describing brain atrophy in MS has generally reported the total volume loss in a given structure, e.g., whole-brain volume decline of −0.7% to −1.0% per year. Only recently has the contribution of normal aging begun to be explored, mainly in the context of developing “pathologic cutoff” values for whole-brain volume loss to discriminate patients with MS from healthy controls (HCs), with the ultimate goal of incorporating such values into a definition of “No Evidence of Disease Activity-4.” However, a systematic, unbiased analysis that considers all brain regions to identify which regions undergo the highest amounts of excess atrophy in MS compared with normal aging has not been undertaken.

The excess atrophy in MS that is beyond normal aging, or “MS-specific atrophy,” may be of fundamental importance because only the excess, disease-related atrophy is presumably modifiable with therapeutic intervention and should represent the therapeutic target for treatments aimed at slowing tissue loss in MS. This requires a robust understanding of normal aging, which is a complex process that varies across structures. After normal head and brain growth during childhood, each brain structure declines with normal aging. The age at which the decline begins varies widely across structures, and the trajectory of decline may be linear or nonlinear depending on the structure. For example, whole-brain atrophy follows a nonlinear trajectory that accelerates significantly with age. Because the rate and trajectory of normal aging decline varies across structures, the amount of MS-specific atrophy should also vary across brain structures and as patients age.

The objectives of this analysis are to (1) identify the top brain regions that undergo the largest, least variable amount of MS-specific atrophy, and (2) test whether the amount of normal aging (and therefore the amount of MS-specific atrophy) increases or decreases in the top regions with age.

Methods

Study population

As described previously, from January 2005 through December 2010, all Caucasian patients aged 18–65 years who fulfilled the 2005 McDonald Criteria for clinically isolated syndrome (CIS) or MS were offered enrollment into a large, prospective phenotype-genotype biomarker study (the EPIC study) at the University of California San Francisco (UCSF) MS Center. More than 500 subjects with CIS/MS were followed longitudinally with annual standardized clinical visits and 3T brain MRIs for up to 5 years. All EPIC participants with relapse-onset MS were included in this analysis.

To increase the HC sample size, 2 HC data sets were pooled for this analysis. The first was a convenience sample of 89 HCs (155 MRI time points) from UCSF, recruited separately from the EPIC cohort, but with brain MRIs performed using the same 3T MRI scanner and protocol as EPIC. In addition, the publicly available Alzheimer’s Disease Neuroimaging Initiative database (ADNI; ida.loni.usc.edu) was queried to select ADNI HCs who had (1) a 3T 3D T1-weighted volumetric sequence, (2) longitudinal data (≥3 time points), and (3) age 55–70 years. This yielded an additional 41 HCs (152 MRI time points), for a total of 130 HCs and 307 MRI time points.

Standard protocol approvals, registrations, and patient consents

Both the EPIC study protocol and the UCSF HC study protocol were approved by the Committee on Human Research at UCSF, and informed consent was obtained from all participants. MRIs from ADNI (ClinicalTrials.gov identifier: NCT00106899) are publicly available.

MRI acquisition

All brain MRI scans from UCSF were acquired using an 8-channel phased array coil in reception and a body coil in transmission on a 3T GE Excite scanner that did not undergo hardware upgrades during the 5-year study period. A 3D, T1-weighted, 1 mm-isotropic, volumetric, inversion recovery spoiled gradient echo (3DT1) sequence (1 × 1 × 1 mm³, 180 slices) was acquired at each time point (Echo Time/Repetition Time/Inversion Time = 2/7/400 milliseconds, flip angle = 15°, 256 × 256 × 180 matrix, 240 × 240 × 180 mm³ field of view, number of excitations = 1).
scanned using a standardized brain MRI protocol at 3T after harmonization and optimization of pulse sequences across centers using a phantom. Per ADNI requirements, MRI scanners must be GE, Siemens, or Phillips systems with a 32-channel coil, and subjects scanned longitudinally at a given site are required to go back to the same scanner, which is an accepted research practice to reduce measurement variability in longitudinal MRI studies.\(^\text{15}\) At each time point, a 3DT1 is acquired in sagittal plane with 1 × 1 × 1.2 mm voxel size. Each scan is reviewed to ensure that the specifications of the ADNI protocol are met, and a single phantom scan is obtained each day that an ADNI participant is scanned. Detailed ADNI procedures and protocols are available online (ida.loni.usc.edu).

**MRI postprocessing**

All 3DT1s were processed using our custom in-house pipeline, which has been described elsewhere\(^\text{8}\) and uses FreeSurfer’s longitudinal processing stream\(^\text{16}\) (v5.3; surfer.nmr.mgh.harvard.edu/fswiki/LongitudinalProcessing), lesion inpainting,\(^\text{17}\) and manual edits as needed. We extracted final output from all available 191 brain regions labeled by FreeSurfer, including subcortical, CSF, ventricular, white matter, and whole-brain volumes, and cortical thickness measurements. In addition, annual percentage whole-brain volume change (PBVC) was calculated from 3DT1s using Structural Image Evaluation Using Normalization of Atrophy\(^\text{18}\) (SIENA; fsl.fmrib.ox.ac.uk/fsl/fslwiki/SIENA).

**Statistical analysis**

**Estimating change over time in each brain region**

All statistical analyses were performed in SAS (v9.4). \(\alpha = 0.05\) was used for all hypothesis testing. Paired left and right volumes were summed, and left and right gyral thickness measurements were averaged, yielding 83 total FreeSurfer regions for analysis. Regional volumes were standardized to percent of intracranial volume. For each brain region, including PBVC from SIENA, percent change from the initial scan was calculated at each time point. The change per year (atrophy slope) in each brain region was estimated in a linear mixed-effects model\(^\text{19}\) as follows:

\[
Y_{ij} = \beta_{00} + \beta_{01}(\text{Time})_{ij} + \beta_{02}MS + \beta_{11}(MS)(\text{Time})_{ij} + b_{0j} + b_{1j}(\text{Time})_{ij} + e_{ij}\]

This model includes fixed effects for on-study time (in years; \(\beta_{01}(\text{Time})_{ij}\)), case-control status (binary; \(\beta_{02}MS\)), the interaction between disease status and time (\(\beta_{11}(MS)(\text{Time})_{ij}\)), which estimates the additional brain atrophy in MS beyond normal aging (i.e., MS-specific atrophy), and random effects for intercept and slope (\(b_{0j}\) and \(b_{1j}(\text{Time})_{ij}\), respectively). Assumptions of the statistical models were checked via residual plots to examine the model integrity, normality of residuals, and constant variance, as well as Cook influence plot to detect potential outliers with a strong influence on model estimates.

**Ranking brain regions**

From the linear mixed-effects model in each region, we extracted the parameter estimate and standard error for the interaction term \(\beta_{11}(MS)(\text{Time})_{ij}\), created a ratio \(\frac{SE(\beta_{11}(MS)(\text{Time})_{ij})}{\beta_{11}(MS)(\text{Time})_{ij}}\) and ranked all 83 FreeSurfer regions according to this ratio. This ratio identifies regions that undergo the highest amount of MS-specific atrophy over time with the least variability in the estimated change (i.e., the lowest standard error). Brain regions were ranked in unadjusted models, after adjusting for age, and after adjusting for both age and sex. Age and sex adjustments were performed by adding age + age*(Time)\(_{ij}\) and sex + sex*(Time)\(_{ij}\) to the model as fixed effects.

**Effect of age on normal aging and MS-specific atrophy**

The effect of age at study entry on normal aging and MS-specific atrophy was explored in the top-ranked FreeSurfer regions and PBVC from SIENA. To test whether normal aging and MS-specific atrophy change with age, age was modeled as an effect modifier in the relationship between MS-specific atrophy and time by adding a 3-way interaction term plus additional marginal terms \(age*MS + age*(Time)_{ij} + age*MS*(Time)_{ij}\) to the model listed above. Using the full model, atrophy slopes for subjects with MS (total atrophy in MS), HCs (normal aging), and the slope difference between MS and HCs (MS-specific atrophy) were estimated at ages 30, 40, 50, and 60 years. These ages were chosen a priori to represent the age range of the subjects with MS in our data set.

**MRI scanner as a potential confounder**

Because our data set was assembled from multiple sources, we tested for potential confounding by testing the interaction between MRI scanner manufacturer/model and atrophy slope (i.e., \(scanner*(Time)_{ij}\)). Visual inspection was used to compare the ranking of brain regions and the point estimate of atrophy slopes at different age categories before and after adding \(scanner*(Time)_{ij}\) to the model.

**Data availability**

Two of the 3 data sets used in this analysis (the EPIC MS data set and the UCSF HC data set) are not sharable publicly. The acquisition of the brain MRI images in the EPIC MS data set was funded by investigator-initiated studies sponsored by pharmaceutical companies. The third data set is publicly available via the ADNI at ida.loni.usc.edu.

**Results**

**Subject characteristics**

Table 1 provides characteristics of the subjects included in this analysis. Six hundred fifty subjects and 2,790 MRI time points were analyzed, including 520 subjects with MS (2,483 MRI time points) and 130 HCs (307 MRI time points). Of the 90 subjects who were classified as CIS at study entry, 47 (52.2%) were reclassified as relapsing-remitting MS (RRMS) or secondary progressive MS (SPMS) by their treating physician during the study period. Of the 392 subjects who were classified as RRMS at study entry, 21 (5.4%) were reclassified as SPMS by their treating physician while on study. Four hundred four subjects (77.7% of the cohort) were exposed to

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disease-modifying therapy (DMT), the vast majority of which were injectable DMTs given the study period (2005–2010).

**Ranking of brain regions**
The top 5 FreeSurfer regions ranked by the slope/standard error ratio are given in table 2. Putamen volume, subcortical gray matter volume, caudate volume, cuneus gyrus thickness, and thalamic volume were consistently ranked as the top 5 regions across all mixed-effects models, including the unadjusted model, after adjusting for age, and after adjusting for age and sex. Examination of residual plots showed a normal distribution of residuals and did not indicate any deviations

| Table 1: Subject demographics and characteristics |
|-----------------------------------------------|
| **Variable** |
| **Subjects with MS (n = 520)** | **HCs (n = 130)** |
| Age at study entry, y | 42.7 ± 9.8 | 47.9 ± 15.1 |
| Age quartiles | | |
| 18–35 | 129 (24.7) | 34 (26.4) |
| 36–43 | 143 (27.6) | 21 (16.3) |
| 44–52 | 142 (27.4) | 22 (17.1) |
| 53–69 | 106 (20.4) | 52 (40.3) |
| Sex | | |
| Female | 365 (70.2) | 83 (63.8) |
| Male | 155 (29.8) | 47 (36.2) |
| Disease course at study entry | | N/A |
| CIS | 90 (17.3) | |
| RRMS | 392 (75.4) | |
| SPMS | 38 (7.3) | |
| Clinical disease duration at study entry, y | 9.2 ± 8.6 | N/A |
| Mean EDSS score at study entry | 1.8 ± 1.5 | N/A |
| Cumulative DMT exposure | | |
| None | 116 (22.3) | |
| ≤1 y | 33 (6.3) | |
| 1–3 y | 77 (14.8) | |
| 3–5 y | 107 (20.6) | |
| >5 y | 187 (36.0) | |
| Mean follow-up time, y | 4.1 ± 1.6 | 1.4 ± 1.5 |
| Total MRI time points analyzed | 2,483 | 307 |
| No. of MRI time points | | |
| Subjects with 1 MRI | 18 | 49 |
| Subjects with 2 MRIs | 47 | 25 |
| Subjects with 3 MRIs | 50 | 24 |
| Subjects with 4 MRIs | 49 | 24 |
| Subjects with 5 MRIs | 111 | 8 |
| Subjects with 6 MRIs | 245 | — |

Abbreviations: CIS = clinically isolated syndrome; DMT = disease-modifying therapy; EDSS = Expanded Disability Status Scale; HC = healthy control; RRMS = relapsing-remitting MS; SPMS = secondary progressive MS.

*Mean ± SD for continuous variables, N (%) for categorical variables.

b Cumulative DMT exposure includes all DMT exposure that occurred prestudy and/or on study for each subject (i.e., throughout their lifetime).
from a linear form, with relatively constant variance across the fitted range. The Cook influence plot did not show any observations with Cook distance >1 to indicate an influential outlier.

**Effect of age on MS-specific atrophy**

The contributions of MS and normal aging to total atrophy in the putamen, caudate, thalamus, and whole-brain volume (PBVC from SIENA) at each decade from age 30 to 60 years are displayed in the figure, with corresponding values in Table 3. “Subcortical gray matter volume” in FreeSurfer includes the thalamus, caudate, hippocampus, amygdala, nucleus accumbens, ventral diencephalon, and substantia nigra (surfer.nmr.mgh.harvard.edu/fswiki/MorphometryStats). Because our aim was to analyze the effect of age on MS-specific atrophy in separate brain regions, subcortical gray matter volume was not retained for further analysis. Cuneus gyral thickness was also not retained because, unlike the deep gray matter nuclei, which are known to be affected by MS,21,22 there is a paucity of previous data suggesting specific involvement of the cuneus gyrus.

As shown in figure A and table 3, the total whole-brain atrophy slope (change/y) in subjects with MS from SIENA was

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**Table 2** Top 5 brain regions ranked by slope/standard error ratio*

| FreeSurfer region                  | Unadjusted | Age adjustedb | Age and sex adjusted |
|------------------------------------|------------|---------------|----------------------|
| Putamen volume                     | −5.05      | −4.74         | −4.68                |
| Subcortical gray matter volume     | −4.11      | −3.44         | −3.44                |
| Caudate volume                     | −4.13      | −3.33         | −3.4                 |
| Cuneus gyrus thickness             | −3.0       | −2.64         | −2.64                |
| Thalamic volume                    | −1.67      | −1.67         | −1.75                |

*a Slope/standard error ratio is as follows: \( \frac{\beta_{MS} \times \text{Time}}{SE} \). The resulting values are unitless.

*b Adjustment was made by adding both the marginal effect and the interaction with time, e.g., age + age × time.

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**Table 3** MS-specific and normal aging atrophy rates by region and decade

| Brain region | Age 30 y       | Age 40 y       | Age 50 y       | Age 60 y       | p Value |
|--------------|----------------|----------------|----------------|----------------|---------|
| SIENA        |                |                |                |                |         |
| Total atrophy (%)/y | −0.37 ± 0.02 | −0.39 ± 0.01  | −0.41 ± 0.02  | −0.43 ± 0.03  |         |
| MS-specific atrophy | −0.38 ± 0.09 | −0.29 ± 0.06  | −0.21 ± 0.05  | −0.12 ± 0.06  | 0.01    |
| Normal aging | 0.01 ± 0.09    | −0.10 ± 0.06  | −0.21 ± 0.05  | −0.31 ± 0.05  | <0.01   |
| Thalamus     |                |                |                |                |         |
| Total atrophy (%)/y | −0.74 ± 0.05 | −0.71 ± 0.03  | −0.69 ± 0.04  | −0.67 ± 0.07  |         |
| MS-specific atrophy | −0.59 ± 0.22 | −0.41 ± 0.15  | −0.23 ± 0.12  | −0.05 ± 0.15  | <0.05   |
| Normal aging | −0.15 ± 0.21   | −0.31 ± 0.15  | −0.46 ± 0.12  | −0.62 ± 0.13  | <0.05   |
| Caudate      |                |                |                |                |         |
| Total atrophy (%)/y | −1.31 ± 0.07 | −1.19 ± 0.04  | −1.07 ± 0.05  | −0.96 ± 0.08  |         |
| MS-specific atrophy | −0.67 ± 0.26 | −0.59 ± 0.19  | −0.52 ± 0.15  | −0.44 ± 0.18  | 0.48    |
| Normal aging | −0.64 ± 0.26   | −0.60 ± 0.18  | −0.56 ± 0.14  | −0.51 ± 0.16  | 0.67    |
| Putamen      |                |                |                |                |         |
| Total atrophy (%)/y | −1.53 ± 0.08 | −1.48 ± 0.05  | −1.42 ± 0.06  | −1.37 ± 0.10  |         |
| MS-specific atrophy | −1.05 ± 0.34 | −0.99 ± 0.24  | −0.93 ± 0.19  | −0.87 ± 0.24  | 0.66    |
| Normal aging | −0.49 ± 0.33   | −0.49 ± 0.23  | −0.50 ± 0.18  | −0.50 ± 0.22  | 0.97    |

Abbreviations: HC = healthy control; SIENA = Structural Image Evaluation using Normalization of Atrophy. Point estimates ± standard error for atrophy rates (percent change per year) in the top brain regions at ages 30, 40, 50, and 60 y for the total rate of atrophy observed in patients with MS (“total atrophy”), the rate of atrophy observed in HCs (i.e., normal aging), and the rate of MS-specific atrophy (slope for MS—slope for HC), estimated with linear mixed-effects models in each region. The p value given refers to the test of difference in normal aging and MS-specific atrophy rates between consecutive decades, with the null hypothesis that the difference between consecutive decades equals zero.
approximately −0.4% per year, consistent across decades. However, the whole-brain atrophy slopes for MS-specific atrophy and normal aging changed significantly with age. The rate of MS-specific whole-brain atrophy decreased from −0.38% per year at age 30 years to −0.12% per year at age 60 years, a change of 0.09% ± 0.035% per decade \((p = 0.01)\), whereas the normal aging atrophy slope increased from 0.01% per year at age 30 years to −0.31% per year at age 60 years, a change of −0.11% ± 0.032% per decade \((p < 0.01)\). A similar pattern was observed in the thalamus (figure, B and table 3).

In the thalamus, the total rate of atrophy in subjects with MS was approximately −0.7% per year, consistent across decades. However, the rate of normal aging atrophy accelerated from −0.15% per year at age 30 years to −0.62% per year at age 60 years \((-0.16% ± 0.079% per decade, p < 0.05)\), whereas the rate of MS-specific atrophy decreased from −0.59% per year at age 30 years to −0.05% per year at age 60 years \((0.18% ± 0.08% per decade, p < 0.05)\).

A very different pattern was observed in the putamen and caudate (figure, C and D and table 3). In these regions, the contributions of normal aging and MS-specific atrophy did
not vary significantly with age. In the caudate, the normal aging and MS-specific atrophy slopes were stable across decades, with nonsignificant changes of 0.04 ± 0.10 (p = 0.67) and 0.07 ± 0.11 (p = 0.48) per decade, respectively. Similarly, in the putamen, the rate of normal aging (~0.01% ± 0.13%, p = 0.97) and MS-specific atrophy (0.06% ± 0.14%, p = 0.66) did not change significantly across decades.

**Type of MRI scanner as a potential confounder**

None of the interaction tests between the type of MRI scanner and the atrophy slope \( (\text{scanner} \times \text{Time}) \) were statistically significant. The rankings and parameter estimates did not change after adding \( \text{scanner} \times \text{Time} \) to the model, suggesting that scanner manufacturer/model did not confound our findings.

**Discussion**

We used an agnostic approach to identify the top brain regions with the highest, least variable amount of yearly MS-specific atrophy in a large, 5-year, prospective cohort of subjects with MS and estimated the contribution of normal aging to atrophy measurements in the top-ranked regions at each decade from ages 30 to 60 years. Out of every region in the human brain that is labeled by FreeSurfer, all the top-ranked regions were gray matter structures, including 3 deep gray nuclei and 1 cortical gyrus. This is consistent with previous histopathologic and MRI studies suggesting involvement of deep and cortical gray matter in the MS disease process\(^{21-23}\) and underscores the importance of gray matter in MS.

More importantly, our findings indicate that for whole-brain (SIENA) and thalamic volume, the contribution of normal aging to total atrophy may increase significantly as the MS population ages. In these regions, most of the atrophy we observed at age 30 years was clearly in excess of normal aging and presumably disease related, whereas by age 60 years, the rate of normal aging had accelerated such that there was little excess atrophy attributable to MS. In the caudate and putamen, on the other hand, we observed a stable contribution of normal aging and MS atrophy has been observed in other studies using SIENA.\(^5\,6\) Although normal aging in subcortical structures is perhaps less well characterized, a similar pattern of behavior could plausibly occur. Some subcortical structures such as the amygdala, caudate, and thalamus have been shown to follow a linear decline over the lifespan in normal aging, whereas others such as the hippocampus may follow a quadratic (i.e., nonlinear) trajectory.\(^{11}\) Most of the studies modeling atrophy in subcortical structures in normal aging have used cross-sectional data sets composed of 1.5T MRIs and relatively simplified statistical methods (i.e., linear or quadratic age terms) to estimate the effect of age.\(^{9-13}\) Our data sets were composed of longitudinal 3T MRIs, and we used a more complex modeling strategy, making direct comparisons with published literature somewhat difficult.

We found that different brain regions behaved differently with respect to MS-specific atrophy vs normal aging. This suggests that tissue loss in MS may occur for reasons that are distinct across brain regions. Axonal transection within white matter demyelinating lesions\(^24\) can lead to upstream or downstream neurodegeneration, which likely contributes to whole-brain and thalamic atrophy in MS.\(^55,56\) This may be a particularly important mechanism of thalamic atrophy, which is an area of early neurodegeneration in MS.\(^26-29\) A strong preponderance for MS lesions to form in thalamocortical white matter and a probable mechanistic link between thalamocortical white matter lesions and thalamic atrophy has been demonstrated using diffusion tensor imaging (DTI) tractography in subjects with CIS.\(^26\) We hypothesize that the downstream effects of white matter lesions may explain the predominantly disease-related atrophy we observed at younger ages in whole brain and thalamus, whereas regions with fewer incoming long white matter tracts such as the caudate and putamen would presumably be less influenced by demyelinating injury. Indeed, in the caudate and putamen, the contributions of normal aging and MS-specific atrophy remained stable with age. Future work could test this hypothesis using a similar DTI tractography approach\(^26\) to delineate incoming and outgoing tracts in the putamen and caudate and to estimate the effect (or lack thereof) of tract-specific white matter lesions on the volume of each structure. If we have indeed identified gray matter regions that are undergoing a high amount of excess MS-specific tissue loss that is relatively independent of white matter lesions, this may open a rich avenue for further exploration that could provide fundamental insights into tissue loss in MS. These regions could be used to study the contribution of other aspects of MS pathology to gray matter atrophy, such as intrinsic gray matter demyelinating lesions,\(^{21,22}\) iron deposition,\(^{22}\) microglial activation,\(^{30}\) mitochondrial dysfunction,\(^{31}\) and glutamate excitotoxicity,\(^{32}\) which could lead to a more comprehensive understanding of gray matter atrophy in MS.

Our findings may have implications for clinical trial design and interpretation. If we presume that only the MS-specific atrophy is amenable to therapeutic intervention (i.e., that DMTs will not slow normal aging), our data suggest that the most appropriate brain region for a phase II primary end point, as well as the effect size that can reasonably be achieved and the sample size needed to demonstrate efficacy, may differ according to the patient population being studied (early vs late MS, older vs younger patients, etc.). More generally, a better understanding of the time and/or age where MS-specific atrophy is most prominent may help elucidate the “therapeutic opportunity” for current and future DMTs. Our findings could at least partly explain the suggestion in our literature that some DMTs aimed at slowing disease progression may not work as well in older patient populations,\(^33-35\) perhaps because normal aging is more prominent and there is
less “therapeutic opportunity” present. Finally, our data support the notion that any “pathologic cutoff” value used to discriminate patients with MS from HCs must be adjusted for age, which has also been suggested by the work of others.5,6

Our findings should not suggest that patients with MS at age 60 years are no longer undergoing pathologic disease-related whole-brain or thalamic atrophy. In fact, the total rate of whole-brain and thalamic atrophy in our subjects was consistent throughout the age range at −0.4% and −0.7% per year, respectively. Rather, our data suggest that normal aging accelerates with age in these regions, which mathematically decreases the excess atrophy because normal aging and MS-specific atrophy were estimated in the same statistical model. Whether whole-brain or regional gray matter atrophy accelerates throughout the MS disease duration has been unclear, with some studies suggesting an acceleration of gray matter atrophy in SPMS compared with RRMS36,37 and other, more recent work finding no difference in deep gray matter atrophy rates across MS subtypes.8,38 We examined the effects of age rather than clinical subtype; although age and clinical subtype are typically correlated, these should not be conflated. Finally, our findings describe a group-level difference between subjects with MS and HCs. Translating these findings to the individual level is an area of high interest that will require the development of methods to create an MS-specific atrophy biomarker for an individual; exploration of the clinical relevance of such a biomarker is likely to be informative. However, presently, it is premature to apply the current findings at the individual level.

Our approach that focuses on excess, disease-specific atrophy could be applied to other imaging metrics in MS and to other neurodegenerative diseases such as Alzheimer disease, Parkinson disease, and Huntington disease, in which a similar conceptual model should apply. If one could estimate and adjust for the impact of normal aging, disease-specific atrophy measurements could allow for comparison across neurodegenerative diseases, which could uncover biologic, environmental, genetic, and phenotypic similarities (and differences) between diseases that could ultimately further our understanding of neurodegeneration in general. Furthermore, an impact of medical and psychiatric comorbidities on brain atrophy has been demonstrated in both MS39–41 and non-MS42–45 subjects. This emerging area deserves further exploration. Because our data sets did not contain information on comorbidities, we were not able to examine their influence on our results. However, all brain regions in a given patient would have been exposed to the same milieu of comorbidities; therefore, it is unlikely that the impact of comorbidities explains our observation of differing behavior between deep gray matter regions. Similarly, the presence or absence of DMT is unlikely to explain our findings, unless one hypothesizes a differential impact of DMT on one deep gray matter structure vs another. Estimating the impact of comorbidities and DMTs on regional brain atrophy in MS will be an area of focus in future studies.

In summary, our findings indicate that the contribution of normal aging may increase with age and the rate of MS-specific atrophy may decrease with age in thalamic and whole-brain atrophy measurements in MS, whereas normal aging and MS-specific atrophy do not change with age in the putamen and caudate. This work emphasizes the need to understand each brain region separately in MS and has many potential implications for MS and perhaps for other neurodegenerative diseases. This could inform future work that furthers our fundamental understanding of brain atrophy in MS, including whether there are different causes of regional tissue loss and how to interpret the effect of disease-modifying therapies with age.

Acknowledgment
The authors thank the research subjects who generously agreed to serve as study participants, the MS clinicians at the UCSF MS Center who referred patients to the EPIC study, the research coordinators at UCSF for data collection, and the ADNI investigators for the creation of the publicly available ADNI database. All authors had full access to the data and take responsibility for the content of this article.

Study funding
This study received support from the NIH National Institute of Neurological Disorders and Stroke (R01NS062885 to

Strengths of this work include the large sample size, longitudinal study design, careful acquisition and postprocessing of MRI data, and the unbiased selection strategy of the top regions of interest, which incorporates normal aging. Our approach of separating normal aging from disease-related atrophy is novel in brain atrophy analyses in the current MS literature and has several implications for MS and other neurodegenerative diseases that could be highly impactful. Limitations include a shorter mean follow-up time in HCs compared with subjects with MS and a relatively small HC sample size. HC MRI data, particularly longitudinal data, are rare in clinical research; efforts to assemble larger HC data sets are ongoing. However, to achieve the meaningful sample sizes that will be needed to fully understand normal aging will invariably require leveraging and combining multiple existing HC data sets composed of subjects scanned on different scanners, as we did in this analysis. Although this may initially prompt concern for increasing measurement variability, each HC subject in this analysis was required to be scanned longitudinally on the same MRI scanner throughout their study period, which reduces variability.15 Moreover, we modeled the brain volume changes in this analysis as a percent change from baseline, which obviates the concern regarding heterogeneity across MRI scanners. Finally, by definition, linear mixed-effects models assume a linear decline in the outcome variable (brain volume) over time. Additional work with larger MRI data sets over a much larger age range is underway and will use advanced, nonlinear modeling strategies to further disentangle the contribution of normal aging from the pathologic atrophy observed in MS.
D.P.) for MRI data processing and Biogen and GlaxoSmithKline for MRI data acquisition. This work was also supported by grant UL1TR001855 from the National Center for Advancing Translational Science (NCATS) of the NIH through an Institutional Career Development Award to C.J.A. The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health.

Disclosure
C.J. Azvedo has received consulting fees from Guerbet, Genentech, Biogen, Novartis, and Sanofi Genzyme. S.Y. Cen, A. Jaberzadeh, and L. Zheng report no disclosures. S.L. Hauser has served on scientific advisory boards for Annexon, Symbiotix, Bionure, and Molecular Stethoscope. He is on the Board of Trustees for Neurona Therapeutics. S.L. Hauser has received travel reimbursement and writing assistance from F. Hoffman-La Roche Ltd. for CD20-related meetings and presentations. D. Pelletier has received consulting and/or speaking fees from Novartis, Roche, Sanofi Genzyme, and Vertex. Go to Neurology.org/NN for full disclosures.

Publication history
Received by Neurology: Neuroimmunology & Neuroinflammation March 16, 2019. Accepted in final form July 1, 2019.

Appendix
Authors

| Name                        | Location       | Role                      | Contribution                                                                 |
|-----------------------------|----------------|---------------------------|-----------------------------------------------------------------------------|
| Christina J. Azvedo, MD, MPH | University of Southern California | Author | Concept and design of the analysis, data interpretation, and drafted a manuscript for intellectual content |
| Steven Y. Cen, PhD          | University of Southern California | Author | Performed the statistical analysis, contributed to the concept and design of this analysis, interpretation of data, and provided critical revision of the manuscript for intellectual content |
| Amir Jaberzadeh, PhD        | University of Southern California | Author | Postprocessing of MRI data                                                  |
| Ling Zheng, PhD             | University of Southern California | Author | Provided critical revision of the manuscript for intellectual content         |
| Stephen L. Hauser, MD       | University of California, San Francisco | Author | Provided critical revision of the manuscript for intellectual content         |
| Daniel Pelletier, MD        | University of Southern California | Author | Concept and design of the analysis, interpretation of data, and provided critical revision of the manuscript for intellectual content |

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