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Association of Gray Matter Atrophy Patterns with Clinical Phenotype and Progression in Multiple Sclerosis

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Abstract

Objectives. Grey matter (GM) involvement is clinically relevant in multiple sclerosis (MS). Using source-based morphometry (SBM), we characterized GM atrophy and its 1-year evolution across different MS phenotypes.

Methods. Clinical and MRI data were obtained at 8 European sites from 170 healthy controls (HCs) and 398 MS patients (34 clinically isolated syndromes [CIS], 226 relapsing-remitting [RR], 95 secondary progressive [SP] and 43 primary progressive [PP] MS). Fifty-seven HC and 144 MS underwent 1-year follow-up. Baseline GM loss, atrophy progression and correlations with disability and 1-year clinical worsening were assessed.

Results. SBM identified 26 cerebellar, subcortical, sensory, motor and cognitive GM components. GM atrophy was found in MS *vs* HC in almost all components (p=range<0.001-0.04). Compared to HCs, CIS patients showed circumscribed subcortical, cerebellar, temporal and salience GM atrophy, while RRMS patients exhibited widespread GM atrophy. Cerebellar, subcortical, sensorimotor, salience and fronto-parietal GM atrophy was found in PPMS patients *vs* HCs, and SPMS *vs* RRMS. At 1-year, 21 (15%) patients had clinically worsened. GM atrophy progressed in MS in subcortical, cerebellar, sensorimotor, and fronto-temporo-parietal components. Baseline higher disability was associated (R²=0.65) with baseline lower normalized brain volume (beta=-0.13, p=0.001), greater sensorimotor GM atrophy (beta=-0.12, p=0.002) and longer disease duration (beta=0.09, p=0.04). Baseline normalized GM volume (odds ratio=0.98, p=0.008) and cerebellar GM atrophy (odds ratio=0.40, p=0.01) independently predicted clinical worsening (area-under-the-curve=0.83). **Conclusion.** GM atrophy differed across disease phenotypes and progressed at 1-year in MS. In addition to global atrophy measures, sensorimotor and cerebellar GM atrophy explained baseline disability and clinical worsening.

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Introduction

In the last decades, grey matter (GM) involvement has been increasingly recognized as a crucial component of pathophysiology in multiple sclerosis (MS).¹ GM atrophy occurs from the earliest MS phases² and progresses over time with rates being 8 times greater than controls in relapsing-remitting MS (RRMS), and up to 14 times greater in progressive MS patients.³ A substantial relationship between GM atrophy and clinical disability has been consistently demonstrated.¹ The regional distribution of GM atrophy is not homogeneous across phenotypes, with an earlier involvement of deep GM and the parietal lobe,^{1, 4, 5} and a progressive spreading of GM loss to frontal, temporal, occipital and cerebellar regions at later disease stages.^{1, 6, 7}

One of the mostly used techniques to investigate GM atrophy localization is voxel-based morphometry,^{8,9} a mass univariate method that does not consider information about the relationships among voxels other than those in the immediate neighborhood. By contrast, independent component analysis (ICA) is able to extract spatially independent sources of GM changes from a set of MR images.¹⁰ In particular, source-based morphometry (SBM)¹¹ is an ICAbased technique that allows to decompose GM maps into distinct patterns of GM density that covary across subjects. These patterns can be thought of as GM "networks" and SBM can be used to quantify between-group differences of GM atrophy regarding such patterns. This is achieved by comparing subject-wise loadings (or weights), which represent the degree to which a pattern is present at an individual subject level. Reduced loadings are usually considered as a measure of atrophy.¹¹ Studies applying SBM in MS found that GM atrophy largely occurs in a non-random manner^{12, 13} and develops in distinct anatomical patterns that show association with clinical disability¹² and, to lesser extent, with white matter (WM) damage.¹³ A 10-year longitudinal SBM study in a cohort of RRMS patients retrospectively extracted from a clinical trial¹⁴ showed that atrophy progression in motor and cognitive GM networks was higher in clinically worsened compared to stable MS patients. Despite these encouraging results, SBM has never been applied to large cohorts of MS patients including all main clinical phenotypes and follow-up evaluations. We

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hypothesize that the use of this technique in such a population would be useful to characterize patterns of atrophy development in the main large-scale GM networks across different disease phenotypes and their relationship with clinical disability and deterioration.

In this study, we applied SBM to a large, multicenter MS dataset, to determine the main patterns of GM atrophy and their 1-year evolution in MS patients compared to healthy controls (HCs), and within the MS population according to their clinical phenotype. We also evaluated the correlation between the occurrence of baseline GM atrophy in specific networks and concomitant clinical impairment, and assessed the predictive role of GM network atrophy on disability worsening.

Methods

<u>Standard Protocol Approvals, Registrations, and Patient Consents</u>. Approval was received from the Local Ethical committees at each participating centre; study participants signed an informed consent prior to enrolment.

Participants. We recruited study subjects at eight European sites from the MAGNIMS consortium (www.magnims.eu): 1) The Amsterdam MS Center, Amsterdam UMC, location VUmc, Amsterdam (The Netherlands); 2) The Cemcat and Section of Neuroradiology, University Hospital Vall d'Hebron, Barcelona (Spain); 3) St. Josef Hospital Ruhr University, Bochum (Germany); 4) Institute of Neurology, UCL, London (UK); 5) The Neurocenter of Southern Switzerland, Lugano (Switzerland); 6) The Neuroimaging Research Unit, San Raffaele Scientific Institute, Milan (Italy); 7) The MRI Center "SUN-FISM", University of Campania "Luigi Vanvitelli", Naples (Italy); 8) The Nuffield Department of Clinical Neurosciences, Oxford (UK). Subjects were part of a previous study aimed at characterizing distribution and regional evolution of cervical cord atrophy.¹⁵

To be included, MS patients had to have stable treatment in the past 6 months and received no corticosteroids during the last month. Patients with a clinically isolated syndrome (CIS) suggestive of MS had to have a first neurological episode suggestive of demyelination and a clinical assessment within three months from symptoms. Patients were excluded if they had history of brain trauma, important comorbidities, drug or alcohol abuse history, or any other medical conditions having an intereference with MRI, inability to undergo MRI (claustrophobia, metal implants, pacemakers, etc.), breastfeeding or pregnancy.

Participants' evaluation included an MRI and clinical assessment at baseline and, whenever possible, after 1-year.¹⁵ In order to exclude increasing disability resulting from relapses close to baseline and follow-up visits, and consequent spurious effects on GM volumes, all patients were free from relapses and steroid treatment for at least one month before clinical and MRI evaluations.

Clinical assessment. Within two days from MRI scanning, MS patients underwent a neurological evaluation by an experienced neurologist, blinded to the MRI results.¹⁵ The Expanded Disability Status Scale (EDSS) score¹⁶ was rated and disease-modifying treatments (DMTs) were recorded. At 1-year follow-up, we considered a patient as clinically worsened if his/her EDSS score increased ≥ 1.0 point when baseline EDSS was <6.0, or if his/her EDSS score increased ≥ 0.5 point when baseline EDSS was $\geq 6.0.^{17}$ EDSS changes were confirmed during a second visit after three-months. If patients did not have confirmed disability progression, they were considered clinically stable.

<u>MRI acquisition</u>. All participating sites used 3.0 T magnets (Siemens Magnetom Trio in Barcelona; Siemens Magnetom Skyra in Lugano; Siemens Magnetom Prisma in Oxford; Philips Achieva in Bochum, Milan and London; General Electric Signa HDtx in Amsterdam and Naples, no scanner upgrades occurring during the study) to acquire the following MRI protocol:¹⁵ a) sagittal brain 3D T1-weighted scan for atrophy assessment (repetition time [TR]=range 6.9-8.58 ms for GE and Philips and 2040-2300 ms for Siemens, echo time [TE]=range 2.8-4.7 ms, inversion time [TI]=range 450-1000 ms, flip angle [FA]=range 8-12°, 128 to 192 sagittal slices with 0.9-1.2 mm slice thickness and \approx 1 mm² in-plane resolution); and b) axial brain dual echo (DE) fast spin echo (TR=range 2500-4670 ms, TE=range 15-27/79-120 ms, FA=range 90°-150°, 44 to 50 axial slices with 3 mm slice thickness and \approx 0.7-0.9 mm² in-plane resolution) or c) 3D sagittal fluid-attenuated

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inversion recovery (FLAIR) (TR/TE=8000/125 ms, TI=2350 ms, FA=variable, 132 sagittal slices, thickness=1.2 mm, in-plane resolution≈1 mm²) for brain lesion assessment.

<u>Conventional MRI analysis</u>. T2-hyperintense and T1-hypointense lesion volumes (LV) of the brain were produced at baseline and follow-up using the Jim 7.0 software package (Xinapse Systems, Colchester, UK). Lesion-filled¹⁸ T1-weighted images were used to calculate baseline normalized brain volume (NBV), normalized GM (NGMV) and normalized WM volumes (NWMV) with FSL SIENAx. Percentage brain volume change at 1-year was calculated with FSL SIENA software.¹⁹ Head size was measured using the inverse of the brain scaling factor derived from FSL SIENAx.

SBM analysis. An optimized, longitudinal pipeline was implemented using a combination of voxel-based⁸ and tensor-based morphometry²⁰ tools, as previously suggested.¹⁴ Briefly, segmentation of baseline and 1-year (when available) 3D T1-weighted images into GM, WM and CSF was performed using SPM12. Subjects having a follow-up MRI evaluation underwent an additional post-processing step, consisting of a non-linear registration of baseline and 1-year T1-weighted scans to generate an unbiased, subject-specific template (SPM12, pairwise registration tool).²⁰ We applied transformations needed to co-register baseline and follow-up images to the corresponding template to native-space GM and WM segmentations. Then, the Diffeomorphic Anatomical Registration using Exponentiated Lie algebra (DARTEL) method was applied to rigidly-aligned GM and WM images of all study subjects, to produce population-specific GM and WM templates and deform images to such templates.⁸ Modulated images were transformed from DARTEL to MNI space using an affine transformation. Finally, GM images were smoothed (8-mm full-width at half maximum Gaussian kernel).

GM maps underwent SBM to produce spatial GM independent components (IC), i.e., groups of spatially distinct GM regions showing common covariations among subjects.¹¹ To this aim, preprocessed baseline and 1-year follow-up GM maps were all concatenated¹⁴ and underwent spatial ICA using the SBM GIFT toolbox (<u>https://trendscenter.org/trends/software/gift</u>) and the Infomax algorithm. The GM matrix was linearly decomposed into a set of ICs, representing patterns of covarying GM volume, and a matrix of loading coefficients, representing the contribution of each scan to a component in terms of GM volume. The model order selected for SBM (i.e., n=98) was determined using the minimum description length criterion. GM loading coefficients were extracted for subsequent statistical analysis. As previously suggested,^{11, 12} GM IC maps and loading coefficients of components with a main negative sign were inverted.

Statistical analysis. SAS version 9.4 was used for all analyses. T2 and T1 LV were logtransformed. Comparisons of demographic, clinical, conventional MRI measures and GM loading coefficients between HC and MS patients were performed generalized linear mixed-effects models adjusted for age, sex and head size. Such models accounted for site heterogeneity and clustering (subjects within sites) using random intercepts. The same models were used for comparisons between MS phenotypes, with the following *post hoc* contrasts, based on disease clinical evolution: HC vs CIS, HC vs primary progressive (PP) MS, CIS vs PPMS, CIS vs RRMS, RRMS vs secondary progressive (SP) MS, and PPMS vs SPMS. Post hoc contrasts were corrected for multiple comparisons using the false discovery rate method.²¹ Changes over time of clinical and MRI variables were compared between HC and MS patients and among phenotypes using generalized linear mixed-effect models adjusted for age, sex, head size and duration of follow-up. Site heterogeneity and clustering was accounted for using random intercepts. In the longitudinal assessment, we grouped together PPMS and SPMS, because of the relatively low number of patients with follow-up assessment. This yielded the following post hoc contrasts: HC vs CIS, CIS vs RRMS and RRMS vs progressive MS patients. HC, clinically worsened and stable MS patients were also compared. To test the effect of DMT, baseline and longitudinal analyses were repeated including in the mixed-effect models the presence of DMT as a binary variable. The significance of the interaction term was also tested.

In MS patients, we tested correlations of EDSS score with disease duration, conventional MRI variables and GM loading coefficients using linear models adjusted for age, sex, site, DMT

and phenotype. Finally, an age-, sex-, site-, DMT- and phenotype-corrected multivariable linear model with a combination of forward and backward stepwise variable selection identified measures independently predicting the EDSS score. The R² index expressed the proportion of variance explained from the model, while the proportional strength of each independent predictor was expressed by standardized coefficients (beta).

Odds ratios (ORs) and related 95% confidence intervals (CI) from logistic regression models were used to assess the association between disease progression and study variables. We adopted a Firth's bias-reduced penalized-likelihood (PL) estimation approach to select the best independent predictors of disease progression using multivariate logistic regression analysis. We conducted forward and backward stepwise analyses, based on PL ratio tests. Confounding covariates included in such regression models were follow-up duration, DMT and centre.

<u>Data Availability</u>. Upon reasonable request, the dataset analyzed in the current study is available from the corresponding author.

Results

Demographic, clinical and conventional MRI assessment. One-hundred and seventy HCs and 398 MS patients were available for the final analysis (Table 1), including 34 CIS, 226 RRMS, 95 SPMS and 43 PPMS. As expected, baseline EDSS (p<0.001) and disease duration (p=0.02) were higher in RRMS *vs* CIS, and in SPMS *vs* RRMS patients (p<0.001). Similarly, brain T2 LV and T1 LV were higher in PPMS and RRMS *vs* CIS patients (p=range <0.001-0.006), and in SPMS *vs* RRMS (p=0.006 and 0.002, respectively) patients. Baseline NBV was lower in PPMS *vs* HC (p<0.001) and SPMS *vs* RRMS (p<0.001). NWMV was not different between MS phenotypes. Between-group conventional MRI differences were not influenced by DMT (p=range 0.10-0.79).

Fifty-seven HCs and 144 MS patients completed the 1-year evaluation (Table 2). The median follow-up duration was 1.01 years (interquartile range [IQR]=0.94-1.07 years) for HCs and 1.08

years (IQR=0.95-1.21 years) for MS patients (p=0.5). Twenty-one (15%) MS patients had clinically worsened at follow-up; 2 patients evolved from CIS to RRMS and 3 evolved from RRMS to SPMS. In MS patients with a follow-up evaluation, EDSS score (p=0.5), T2 LV (p=0.07) and T1 LV (p=0.43) did not change over time, and percentage brain volume change did not differ between HC and MS patients (Table 2). Clinically stable and worsened MS differed in terms of sex (p=0.01), baseline EDSS (p=0.05), T2 LV (p=0.04) and percentage brain volume change (p=0.02), but they did not differ in terms of age (p=0.4), disease duration (p=0.2) and T1 LV (p=0.06).

<u>Baseline SBM analysis</u>. Among the 98 ICs estimated by SBM, 26 relevant GM components were selected by visual inspection, according to their correspondence with well-known sensorimotor and cognitive networks.^{12, 14} IC provided by Allen et al.²² were also used as reference templates, given the similarity of model order between this study (n=100) and our study. Relevant GM IC (thresholded using a Z-score>2.5 and a cluster extent >100 mm³) were assigned to the following networks: cerebellar (4 IC), subcortical (3 IC), sensorimotor (4 IC), visual (2 IC), auditory (2 IC), default-mode (DMN, 4 IC), fronto-parietal (5 IC) and salience (2 IC). Illustrative examples of such IC are shown in Figure 1; while a complete, quantitative description of selected IC is available on Dryad (Figure e-1) https://doi.org/10.5061/dryad.mcvdncjzx. Compared to HC, MS patients exhibited significant GM atrophy in all IC (p=range <0.001-0.036), except for one (p=0.11) in the fronto-parietal network and one (p=0.07) in the DMN (Table 3).

A complete description of baseline GM atrophy comparisons among phenotypes is reported in Table 3 and Figure e-2 (available on Dryad) https://doi.org/10.5061/dryad.mcvdncjzx. Illustrative snapshots of such comparisons are also shown in Figure 2. Compared to HC, CIS patients showed circumscribed GM atrophy in one cerebellar, two subcortical, and one auditory IC (p=range <0.001-0.04, Table 3, Figure 2). RRMS patients exhibited widespread GM atrophy *vs* HC in most of GM IC (p=range <0.001-0.05, Table 3, Figure 2), and additional GM involvement *vs* CIS in one subcortical and one DMN IC (p=0.001 and 0.04 respectively, Table 3). PPMS patients showed a pattern of GM atrophy *vs* HC that was mainly located in cerebellar, subcortical, sensorimotor,

visual, auditory, fronto-parietal, DMN and salience IC (p=range <0.001-0.02, Table 3, Figure 2). The same IC showed additional GM atrophy in SPMS *vs* RRMS, except for the cerebellar network (p=range <0.001-0.04, Table 3, Figure 2). The direct comparison of the two progressive phenotypes showed more GM atrophy in subcortical IC in SPMS *vs* PPMS (p=range <0.001-0.01, Table 3) and more GM atrophy in one sensorimotor IC in PPMS *vs* SPMS (p<0.001, Table 3). Higher atrophy in two sensorimotor ICs in SPMS *vs* RRMS patients was significantly influenced by DMT, with larger GM volume loss detected in untreated than in treated patients (p=0.02 and 0.04, respectively). DMT did not have any effect on the remaining comparisons (p=range 0.07-0.91).

<u>Longitudinal analysis</u>. GM volume remained stable in HCs. Conversely, GM atrophy significantly progressed in MS patients in 5 ICs, including a cerebellar, a subcortical, a sensorimotor, an auditory and a fronto-parietal IC (p=range <0.001-0.04, Table 4, Figure 3). No DMT *x* atrophy progression interaction was detected (p=range 0.14-0.92).

Longitudinal GM volume change was mainly driven by progressive MS, which showed significantly decreased GM volume in the cerebellar (p=0.02), sensorimotor (p=0.02), auditory (p=range < 0.001-0.02), fronto-parietal (p=0.03), and salience (p=0.02) networks (Figure 3). The sensorimotor network showed a significant GM decrease over time also in RRMS (p=0.03) and CIS (p=range 0.02-0.04) patients. Finally, in CIS patients, a significant GM atrophy development was found in the cerebellar (p=0.02) network (Figure 3). No influence of DMTs on GM atrophy progression within phenotypes was detected (p=range 0.21-0.98).

In both clinically stable and worsened MS patients, a significant GM decrease over time was detected in the cerebellar (p=0.02 and 0.01, respectively), sensorimotor (p=0.002 and 0.01, respectively) and auditory (p=0.03 and 0.02, respectively) networks, with no significant time *x* group nor DMT x group interactions (p=range 0.14-0.97). Clinically stable MS showed significant GM loss in the fronto-parietal network (p=0.006).

<u>Correlation analysis</u>. A higher baseline EDSS score was significantly correlated, in MS patients, with longer disease duration (beta=0.13, p=0.008), higher brain T2 LV (beta=0.12,

p=0.002) and T1 LV (beta=0.09, p=0.01) and lower NBV, NWMV and NGMV (beta range=-0.13/-0.18, p<0.001). In univariate analyses, a higher EDSS was significantly associated with GM atrophy in several SBM-derived IC, mainly belonging to the subcortical, sensorimotor, cerebellar, fronto-parietal and salience networks (beta=range -0.07/-0.17, p=range <0.001-0.03). In multivariate analyses, a higher baseline EDSS score was associated (R^2 =0.65) with lower NBV (beta=-0.13, p=0.001), higher GM atrophy in the sensorimotor network (beta=-0.12, p=0.002) and longer disease duration (beta=0.09, p=0.04).

The univariate logistic regression models showed that male sex (OR=3.2, 95% CI=1.2-11.7, p=0.02), higher T2 (OR=2.5, 95% CI=1.1-6.3, p=0.03) and T1 (OR=2.7, 95% CI=1.0-5.4, p=0.04) LV, and lower NBV (OR=0.99, 95% CI=0.98-1.0, p=0.04), NGMV (OR=0.98, 95% CI=0.97-0.99, p=0.01) and GMV in the cerebellar network (OR=0.47, 95% CI=0.24-0.99, p=0.02) were significantly associated with disability worsening. The stepwise multivariable logistic model retained lower NGMV (OR=0.98, 95% CI=0.96-0.99, p=0.008) and lower GMV in the cerebellar network (OR=0.40, 95% CI=0.19-0.85, p=0.01) as independent predictors of disability worsening (area under the curve=0.83).

Discussion

Although MS is traditionally considered a WM disease, GM involvement has been recognized as a crucial determinant of clinical manifestations and prognosis.¹ Trajectories of GM atrophy development in MS are not completely understood, and a variable number of cortical and subcortical GM structures have been shown to be affected.¹ Previous studies substantially agree in showing that GM atrophy is clinically relevant, being useful in characterizing the main disease clinical phenotypes,²³ and being able to explain specific disease-related symptoms, disability progression and cognitive deficits.^{1, 5, 7}

Results of studies applying voxel-based morphometry are not fully consistent, and may be limited by the small sample sizes or by the inclusion of a single disease phenotype.^{6, 24-27} In this

work, a rather novel multivariate method (i.e., SBM) was applied to identify spatial patterns of covarying GM volume in patients with MS. Such an analysis identified 26 distinct, non-random GM IC, which ranged from patterns associated with motor or sensory related areas, to subcortical or cerebellar patterns, and to regions associated with cognition (DMN, fronto-parietal and salience networks). This is in line with preliminary SBM studies in MS, which were, however, limited by the lack of a follow-up examination,^{12, 13} or by the absence of reference HCs and by the inclusion of a single disease phenotype.¹⁴ Our components were less spatially extended than those of the previous studies, ¹²⁻¹⁴ each covering small, but homogeneous brain portions. This is probably due to our relatively high ICA model order (n=98). However, model order selection is likely to play a minor influence on results, since SBM findings were consistent across a variety of different settings.²⁸

Compared to HCs, MS patients exhibited significant baseline GM atrophy in almost all IC. This is in line with previous volumetric studies^{5, 6, 24-27} and confirms that there is a strong neurodegenerative component in MS leading to irreversible tissue loss.¹ Atrophy started to occur relatively early in the disease, since CIS patients already showed significantly GM atrophy *vs* HCs in subcortical and cerebellar IC and a marginal involvement of the temporal lobe. This confirms the early vulnerability to damage of subcortical GM,^{4, 29} which might be due its proximity to the ventricular system-mediated pathogenetic factors.³⁰⁻³²

RRMS patients exhibited a widespread pattern of GM atrophy *vs* HCs, involving all subcortical, cerebellar and sensorimotor ICs, as well as some cognitive, higher-order IC. This suggests a progressive spreading of GM pathology from subcortical to cortical areas, especially those involved in sensory and motor functions. In the same networks, diffuse GM loss became more severe in SPMS *vs* RRMS patients.^{1, 5, 24} In line with previous work, PPMS patients showed GM atrophy *vs* HCs mainly in sensorimotor, auditory, and subcortical networks.²⁶ In addition, we also found GM atrophy in cerebellar, visual, fronto-parietal and salience networks. Taken together, these results suggest that SBM may be sensitive to the heterogeneity of MS-related structural damage,

being powerful in detecting between-group differences of GM atrophy among different MS phenotypes.^{24-27, 29, 33}

At 1-year follow-up, GM atrophy significantly progressed over time in MS patients in 5 IC, located in distinct anatomical systems. This supports the notion that GM atrophy progression in MS patients is extended and severe, and that the neurodegenerative process related to MS may lead to a premature brain aging.³⁴ Progression of GM atrophy was mainly driven by progressive MS patients, who exhibited significant cerebellar, sensorimotor, fronto-parietal and temporal GM atrophy. This is in line with studies demonstrating that brain atrophy accelerates in the progressive disease phase.^{3, 25} On the other hand, atrophy significantly progressed over time also in CIS patients especially in the subcortical compartment, reinforcing the notion of a preferential damage of deep GM and cerebellar structures at the initial disease stages.^{4, 29}

Interestingly, clinically worsened and stable MS patients had a similar pattern of GM atrophy evolution. This seems to be partially counterintuitive and in contrast with previous studies, which detected a higher cortical atrophy progression in clinically worsened than in stable patients.^{7, 14, 35-37} Several factors could explain the lack of differences between stable and worsened patients, including the small number of patients with follow-up, the relatively small number of patients with clinical deterioration during the relatively short duration of the 1-year follow-up (which might not be sufficient to detect clinically relevant modifications), and the subtle contribution of relapses between baseline and follow-up to disability accumulation. Considering that brain atrophy represents an end-stage phenomenon that could develop and progress also several months after the accumulation of focal lesions due to a secondary retrograde degeneration, further studies with longer follow-up are needed to better assess possible differences between these two groups.

In MS patients, a higher baseline EDSS score was significantly associated with GM atrophy of several components, belonging to all main motor, sensory and high-order networks. This suggests that the multiparametric evaluation of atrophy of several systems is able to explain overall disease severity.²⁷ Moreover, the multivariate analysis found that the variables best associated with a higher EDSS score were a lower NBV, higher GM atrophy in the sensorimotor network and longer disease duration. This indicates that GM atrophy of the sensorimotor network could be crucial to explain MS-related clinical impairment. This is not unexpected, given that the EDSS score is heavily weighted towards locomotor dysfunction.

Finally, using univariate logistic regression models, male sex, higher lesion volumes, lower NBV, NGMV and lower GM volume in one cerebellar IC predicted 1-year disability worsening. The stepwise multivariable logistic model confirmed that lower NGMV and lower GM volume in the cerebellar SBM network predicted 83% of disability worsening over 1-year. These results are in line with previous studies regarding biomarkers for disability prediction. For instance, males with MS were shown to have a worse prognosis and a more aggressive MS course than females.³⁸ In other studies, baseline GM damage predicted long-term disability and cognitive deterioration at 13-year¹⁷ and 15-year³⁹ follow-up. Lastly, in MS patients, lower cerebellar volumes were associated with poor motor and cognitive performance in a cross-sectional study.⁴⁰ Such results might reflect the extensive projections to the cerebellum from the limbs via the ventral and dorsal spino-cerebellar tracts, and from motor cortices through the middle cerebellar peduncles.⁴¹

This study has some limitations. First, despite the large number of participants recruited, CIS and PPMS patients were less numerically represented than other phenotypes, probably reflecting their smaller prevalence in the general MS population. Moreover, the number of progressive MS patients who completed the follow-up assessment was relatively small; therefore, we could not reliably calculate GM atrophy progression for PPMS and SPMS phenotypes, separately. Third, MS patients were older than HCs, possibly causing an enhancement of differences in GM volume between these two groups. However, statistical age adjustment was included in all statistical models. Fourth, the large majority of RRMS patients received a DMT, while the opposite was true for progressive MS patients. This numerical imbalance is probably the reason why we did not detect a significant influence of DMT on GM atrophy progression within phenotypes. Fourth, clinical worsening was assessed on a relatively short duration of follow-up (i.e.,

17

1 year), which might not be sufficient to detect clinically relevant modifications in the majority of patients. Finally, neuropsychological assessment was not performed in our patients, preventing us from investigating correlations with cognitive impairment, depression or fatigue symptoms.

To conclude, SBM analysis revealed a differential involvement in various GM networks across disease stages, which progressed at 1-year in MS. Sensorimotor and cerebellar GM atrophy explained baseline disability and clinical worsening.

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controls (HCs) and	patients w	ith multiple	e sclerosis	s (MS) wh	o complete	d the base	line evalu	ation.
MS patients are first	st considere	ed as a who	le, and th	en divideo	l according	to their c	linical phe	enotype.
Study subjects	HCs	MS	p*	CIS	RRMS	SPMS	PPMS	p*
at baseline	(n=170)	(n=398)		(n=34)	(n=226)	(n=95)	(n=43)	
M/F	68/102	163/235	0.85	13/21	84/142	40/55	26/17	0.1
Mean age [y]	40.0	46.8	< 0.001	33.8	44.0	54.3	55.4	< 0.001
(range)	(19-75)	(18-77)		(19-50)	(18-70)	(33-72)	(27-77)	
Mean disease	-	14.1	-	0.5	12.7	21.5	15.8	< 0.001
duration [y]		(0.1-46)		(0.08-	(0.1-37)	(3-46)	(2-45)	
(range)				3)				
Median baseline	-	3.5	-	1.5	2.5 (0.0-	6.0	6.0	< 0.001
EDSS (range)		(0.0-8.5)		(0.0-	6.5)	(2.5-	(3.0-	
				4.0)		8.0)	8.5)	
N° (%) of	-	242	-	4	205	23	10	< 0.001
patients		(61%)		(12%)	(91%)	(24%)	(23%)	
receiving								
disease-								
modifying								
treatments								
Mean brain T2	0.02	10.4	< 0.001	2.4	8.4 (9.8)	18.3	9.6	< 0.001
LV [ml] (SD)	(0.9)	(12.4)		(4.2)		(16.1)	(10.8)	
Mean brain T1	-	6.7	-	2.0	5.3 (6.6)	11.7	7.1	< 0.001
LV [ml] (SD)		(9.0)		(3.7)		(12.7)	(8.4)	
Mean NBV [ml]	1478	1416	< 0.001	1512	1426	1369	1387	< 0.001
(SD)	(65)	(83)		(86)	(75)	(68)	(66)	
Mean NGMV	789 (48)	751	< 0.001	799	760 (50)	727	715	< 0.001
[ml] (SD)		(52)		(51)		(41)	(36)	

Table 1. Main demographic, clinical and conventional baseline MRI characteristics of healthy 1 (770) ad a stigate with a o a man lata ditha haa alima ay

*linear mixed effect model, accounting for clustering (subjects within sites).

< 0.001

712

(50)

665 (47)

641

(49)

671

(42)

664

(50)

Mean NWMV

[ml] (SD)

689 (39)

< 0.001

Abbreviations: SD=standard deviation; CIS=clinically isolated syndrome; RR=relapsingremitting; SP=secondary progressive; PP=primary progressive; EDSS=Expanded disability status scale; LV=lesion volume; NBV=normalized brain volume; NGMV=normalized grey matter volume; NWMV=normalized white matter volume. **Table 2.** Main demographic, clinical and conventional MRI characteristics of healthy controls (HCs) and patients with multiple sclerosis (MS) who completed the follow-up assessment. MS patients are first considered as a whole, and then divided into clinically worsened and clinically stable.

Study subjects (longitudinal)		HCs (n=57)	MS (n=144)	p*	Clinically stable MS (n=122)	Clinically worsened MS (n=21)	p**
M/F		24/33	66/78	0.64	50/73°°	16/5°	0.01
Mean age [y] (range)		35.5 (19-75)	41.5 (18-72)	<0.001	40.6 ⁺ (18-67)	45.8 ⁺ (27-72)	< 0.001
Mean baseline disease duration [y] (range)		-	9.4 (0.08- 42)	-	9.3 (0.08-42)	9.5 (0.25-31)	0.2
Median follow-up duration [y] (interquartile range)		1.01 (0.94- 1.07)	1.08 (0.95- 1.21)	0.5	1.13 (0.99-1.20)	1.29 (0.97-1.5)	0.15
Median EDSS score	Baseline	-	2.5 (0.0- 7.5)	-	2.0 ^{°°} (0.0-7.5)	5.0° (0.0-6.5)	0.05
(range)	Follow- up	-	2.5 (0.0- 8.0)	-	2.0°° (0.0-7.5)	6.0° (1.0-8.0)	< 0.001
Mean brain T2 LV [ml]	Baseline	0.02 (0.9)	6.7 (8.6)	< 0.001	5.8 ⁺⁰⁰ (7.8)	10.8 ^{+°} (10.9)	< 0.001
(SD)	Follow- up	0.02 (0.8)	6.9 (8.3)	< 0.001	6.3 ⁺⁰⁰ (7.3)	10.6 ⁺ ° (11.0)	< 0.001
Mean brain T1 LV [ml]	Baseline	-	5.3 (7.4)	-	4.4 (6.1)	9.8 (11.0)	0.06
(SD)	Follow- up	-	5.5 (7.4)	-	4.8 (6.3)	9.6 (11.1)	0.16
PBVC [%] (SD)		0.02 (0.54)	-0.38 (1.07)	0.08	-0.28 °° (1.07)	-1.00 ⁺⁰ (0.93)	0.01

Linear mixed effect model, accounting for clustering (subjects within sites): *difference between HCs and all MS; **global heterogeneity among HCs, clinically stable and worsened MS: +significant *vs* HCs; °significant *vs* clinically stable MS; °°significant *vs* clinically worsened MS

Abbreviations: SD=standard deviation; EDSS=Expanded Disability Status Scale; LV=lesion volume; PBVC=percentage brain volume change.

Table 3. Average baseline loading coefficients of relevant grey matter (GM) independent

 components (IC) derived from the source-based morphometry (SBM) analysis in healthy controls

 (HCs) and patients with multiple sclerosis, first considered as a whole and then divided into

 different clinical phenotypes.

		MS		CIS	RRMS	SPMS	PPMS	
	HCs	patients		patients	patients	patients	patients	
Network	Mean	Mean	p*	Mean	Mean	Mean	Mean	p**
	(SD)	(SD)		(SD)	(SD)	(SD)	(SD)	
Cerebellar	0.40	-0.13	< 0.001	-0.04	-0.12	-0.19	-0.15	< 0.001
(IC #01)	(0.91)	(0.84)		(0.74)+	(0.85)+	(0.87)	(0.84)	
Cerebellar	0.45	-0.025	< 0.001	-0.11	0.05	0.01	-0.25	< 0.001
(IC #09)	(0.94)	(0.98)		(0.92)	(0.95)+	(1.07)	(0.89)+	
Cerebellar	0.45	-0.088	< 0.001	-0.11	0.04	-0.24	-0.27	< 0.001
(IC #13)	(0.90)	(0.99)		(1.00)	(1.04)+	(0.92)	(0.68)	
Cerebellar	0.26	-0.11	0.003	0.16	-0.09	-0.19	-0.20	< 0.001
(IC #39)	(1.02)	(0.92)		(1.03)	(0.92)+	(0.94)	(0.72)	
Subcortical	0.78	-0.37	< 0.001	0.26	-0.29	-0.94	-0.02	< 0.001
(IC #02)	(0.61)	(0.97)		(0.62)+	(0.91)+	(0.89) ^{§/°}	(1.01)*	
Subcortical	0.53	-0.30	< 0.001	0.51	-0.23	-0.85	-0.07	< 0.001
(IC #10)	(0.73)	(1.02)		(0.59)	(0.95) +/++	(1.10) ^{§/°}	(0.87)*	
Subcortical	0.61	0.021	< 0.001	0.10	0.08	-0.13	0.11	< 0.001
(IC #14)	(0.94)	(0.94)		(0.97)*	(0.95)+	(0.95) °	(0.84)	
Sensorimotor	0.26	-0.13	0.002	0.13	-0.08	-0.21	-0.45	< 0.001
(IC #06)	(0.93)	(0.96)		(0.84)	(0.91)+	(1.11)	(0.92)+	
Sensorimotor	0.38	-0.17	< 0.001	0.35	-0.09	-0.58	-0.08	< 0.001
(IC #34)	(0.83)	(1.02)		(0.91)	(0.98)+	(0.96)	(1.17)	
Sensorimotor	0.54	-0.085	< 0.001	0.11	-0.06	-0.13	-0.28	< 0.001
(IC #53)	(0.93)	(0.92)		(0.67)	(0.93)+	(1.04)§	(0.69)	
							+/00	
Sensorimotor	0.44	-0.18	< 0.001	0.24	-0.05	-0.59	-0.33	< 0.001
(IC #74)	(0.82)	(0.98)		(0.73)	(0.91)+	(1.04)	(1.01)	

Visual	0.35	-0.26	< 0.001	0.37	-0.12	-0.79	-0.38	< 0.001
(IC #40)	(0.86)	(0.97)		(0.69)	(0.93)+	(0.92)§	(1.09)+	
Visual	0.12	-0.22	0.001	0.14	-0.16	-0.50	-0.20	< 0.001
(IC #51)	(0.98)	(0.99)		(0.79)	(0.87)+	(1.19)	(1.05)*	
Auditory	0.49	-0.20	< 0.001	-0.03	-0.03	-0.59	-0.39	< 0.001
(IC #42)	(0.97)	(0.96)		(0.87)+	(0.94)+	(0.86)§	(1.04)+	
Auditory	0.29	-0.20	< 0.001	0.05	-0.05	-0.58	-0.33	< 0.001
(IC #78)	(0.91)	(0.94)		(0.70)	(0.94)+	(0.88)	(0.95)	
Default-mode	0.32	-0.20	< 0.001	0.55	-0.15	-0.56	-0.28	< 0.001
(IC #05)	(0.91)	(0.97)		(0.95)	(0.94) +/++	(0.95)	(0.78)	
Default-mode	0.33	-0.21	< 0.001	0.29	-0.18	-0.55	-0.07	< 0.001
(IC #07)	(0.91)	(0.98)		(1.04)	(0.98)+	(0.85)	(1.03)	
Default-mode	0.16	-0.095	0.07	0.30	0.02	-0.55	-0.09	< 0.001
(IC #27)	(0.90)	(0.99)		(0.92)	(0.93)	(1.04)	(0.89)	
Default-mode	0.22	-0.21	< 0.001	-0.08	-0.11	-0.55	-0.01	< 0.001
(IC #31)	(0.81)	(0.77)		(0.54)	(0.72)+	(0.85)§	$(0.84)^{+}$	
Fronto-	0.30	-0.079	0.02	0.24	0.04	-0.35	-0.37	< 0.001
parietal (IC	(0.97)	(0.96)		(0.92)	(0.95)	(1.01)	(0.69)	
#15)								
Fronto-	0.29	-0.22	< 0.001	0.35	-0.02	-0.67	-0.74	< 0.001
parietal (IC	(0.90)	(0.98)		(0.83)	(0.96)*	(0.82)§	(0.86)*	
#23)								
Fronto-	0.27	-0.038	0.20	0.08	0.10	0.05	-0.37	< 0.001
parietal (IC	(0.98)	(0.94)		(0.89)	(0.91)	(1.04)	(0.80)	
#46)								
Fronto-	0.42	-0.23	< 0.001	0.36	-0.09	-0.71	-0.39	< 0.001
parietal (IC	(0.98)	(0.97)		(0.86)	(0.91)+	(0.93)	(0.92)+	
#49)								
Fronto-	0.33	-0.23	< 0.001	0.16	-0.05	-0.66	-0.48	< 0.001
parietal (IC	(0.94)	(0.88)		(0.82)	(0.86)+	(0.73) [§]	(0.87)+	
#95)								
Salience	0.53	-0.33	< 0.001	0.36	-0.17	-0.82	-0.61	< 0.001
(IC #61)	(0.96)	(0.90)		(0.65)	(0.88)+	(0.77) [§]	(0.81)+	

Salience	0.33	-0.16	0.001	0.42	0.02	-0.63	-0.53	< 0.001
(IC #97)	(0.88)	(0.94)		(0.80)	(0.94)	(0.77) [§]	(0.81)	

Linear mixed effect model, accounting for clustering (subjects within sites): *difference between HCs and all MS; **global heterogeneity among phenotypes; *post hoc* comparisons: *significant *vs* HCs; **significant *vs* CIS; [§]significant *vs* RRMS; °significant *vs* PPMS; °°significant *vs* SPMS.

Abbreviations: CIS=clinically isolated syndrome; RR=relapsing-remitting; SP=secondary progressive; PP=primary progressive

Table 4. Average baseline and follow-up loading coefficients of relevant grey matter (GM)

 independent components (IC) derived from the source-based morphometry (SBM) analysis in

 patients with multiple sclerosis (MS) and healthy controls (HCs) who completed the follow-up

 assessment.

Network	Baseline	Follow-up	p*	Baseline MS	Follow-up	p*	p**
	HCs Mean	HCs Mean		patients	MS		
	(SD)	(SD)		Mean (SD)	patients		
					Mean (SD)		
Cerebellar	-	I I			I		
IC #01	0.22 (0.96)	0.18 (1.01)	0.08	-0.20 (0.80)	-0.21	0.24	0.23
					(0.79)		
IC #09	0.03 (0.88)	-0.02 (0.97)	0.86	-0.42 (0.87)	-0.47	0.10	0.38
					(0.84)		
IC #13	0.41 (0.97)	0.35 (1.01)	0.59	-0.38 (0.90)	-0.43	0.004	0.22
					(0.88)		
IC #39	0.16 (1.12)	0.09 (1.27)	0.42	-0.02 (0.97)	-0.04	0.51	0.65
					(1.00)		
Subcortical	 [I I			I		
IC #02	0.80 (0.63)	0.78 (0.71)	0.73	-0.22 (0.78)	-0.22	0.81	0.85
					(0.79)		
IC #10	0.73 (0.77)	0.72 (0.79)	0.78	-0.07 (0.90)	-0.09	0.26	0.07
					(0.88)		
IC #14	-0.19	-0.20 (0.84)	0.95	-0.66 (0.77)	-0.70	0.04	0.24
	(0.83)				(0.77)		
Sensorimot	tor	I I			I		
IC #06	0.39 (1.01)	0.38 (1.04)	0.16	-0.11 (0.93)	-0.14	0.11	0.59
					(0.91)		
IC #34	0.34 (0.79)	0.34 (0.81)	0.70	-0.09 (1.02)	-0.12	0.71	0.90
					(1.03)		
IC #53	-0.02	-0.09 (0.83)	0.07	-0.38 (0.82)	-0.42	0.27	0.22
	(0.84)				(0.83)		

IC #74	0.37 (0.94)	0.32 (0.91)	0.30	-0.13 (0.92)	-0.18	< 0.001	0.32
					(0.93)		
Visual		I I			I		
IC #40	0.54 (0.74)	0.55 (0.73)	0.59	0.04 (0.98)	0.02 (0.96)	0.13	0.69
IC #51	0.54 (0.96)	0.52 (1.01)	0.46	0.25 (0.91)	0.23 (0.89)	0.12	0.73
Auditory					I		
IC #42	0.51 (0.90)	0.49 (0.92)	0.35	-0.17 (0.91)	-0.19	0.03	0.58
					(0.91)		
IC #78	0.65 (1.02)	0.61 (1.02)	0.06	-0.04 (0.95)	-0.07	0.08	0.46
					(0.96)		
Default-mo	ode				I		
IC #05	0.45 (1.02)	0.44 (1.03)	0.32	0.005 (0.98)	-0.04	0.06	0.95
					(1.00)		
IC #07	0.57 (0.93)	0.57 (0.92)	0.95	-0.02 (0.96)	-0.04	0.34	0.14
					(0.96)		
IC #27	0.20 (1.10)	0.19 (1.01)	0.36	-0.03 (0.99)	-0.05	0.29	0.72
					(0.98)		
IC #31	0.50 (0.80)	0.48 (0.78)	0.11	0.01 (0.69)	-0.01	0.28	0.53
					(0.68)		
Fronto-pari	etal						
IC #15	0.07 (1.09)	0.05 (1.01)	0.32	-0.16 (0.98)	-0.19	0.007	0.93
					(1.01)		
IC #23	0.43 (0.94)	0.42 (0.96)	0.50	0.08 (1.03)	0.06 (1.02)	0.23	0.69
IC #46	-0.27	-0.30 (0.94)	0.11	-0.34 (0.87)	-0.36	0.38	0.28
	(0.97)				(0.89)		
IC #49	0.54 (0.93)	0.51 (0.92)	0.37	-0.05 (0.92)	-0.08	0.29	0.92
					(0.91)		
IC #95	0.48 (1.11)	0.46 (1.11)	0.56	-0.005 (0.98)	-0.01	0.75	0.68
					(0.97)		
Salience	1						
IC #61	0.78 (0.83)	0.74 (0.82)	0.11	-0.03 (0.80)	-0.07	0.08	0.86
					(0.81)		

IC #97	0.20 (0.93)	0.14 (0.93)	0.11	-0.03 (1.09)	-0.06	0.14	0.46
					(1.10)		

*Linear mixed effect model accounting for clustering (subjects within sites) and for repeated measures: *within-group effect of time, **time by group interaction.

Abbreviations: SD=standard deviation; IC=independent component

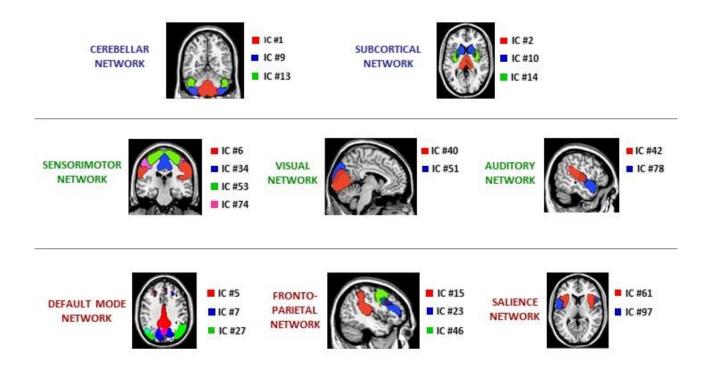


Figure 1. Illustrative map of relevant independent components (IC) obtained after the selection procedure. After running source-based morphometry with n=98 components, relevant IC were selected and sorted into eight subcategories: cerebellar, subcortical, sensorimotor, visual, auditory, default-mode, fronto-parietal, and salience networks. Each color in the composite map corresponds to a different IC within a given subcategory. IC patterns were thresholded at Z>2.5. Images are in neurological convention.

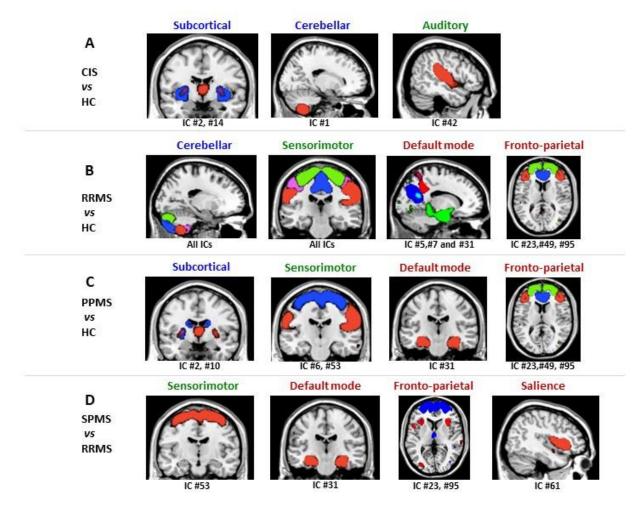


Figure 2. Illustrative map of relevant independent components (IC) showing significant baseline differences of grey matter atrophy between phenotypes at *post hoc* **analysis.** A) clinically isolated syndrome (CIS) patients *vs* healthy controls (HC); B) RRMS patients *vs* HC; C) primary progressive multiple sclerosis (PPMS) patients *vs* HC; and D) secondary progressive multiple sclerosis (SPMS) *vs* RRMS patients. Each color in the composite map corresponds to a different IC within a given subcategory. Images are in neurological convention.

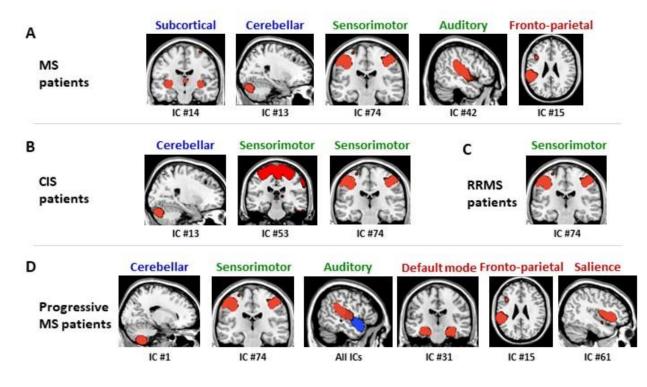
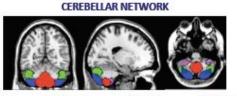
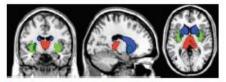


Figure 3. Composite map of the relevant independent components (IC) showing significant progression of grey matter atrophy at 1-year of follow-up. A) Grey matter loss over time in multiple sclerosis (MS) patients, considered as a whole; B) Grey matter loss over time in clinically isolated syndrome (CIS) patients; C) Grey matter loss over time in relapsing-remitting multiple sclerosis (RRMS) patients; and D) grey matter loss over time in progressive MS patients. Images are in neurological convention.



IC #1 IC #9 IC #13 IC #39 IC #39

SUBCORTICAL NETWORK



E IC #2 IC #10 IC #14 L Thai (6986 mm³) L Putam (773 mm³) L Caud (6574 mm³) L Pall (867 mm³) L Putam (6571 mm³) R Thai (6675 mm³) R Putam (1073 mm³) R Caud (7448 mm³) R Pall (779 mm³) R Putam (7185 mm³)

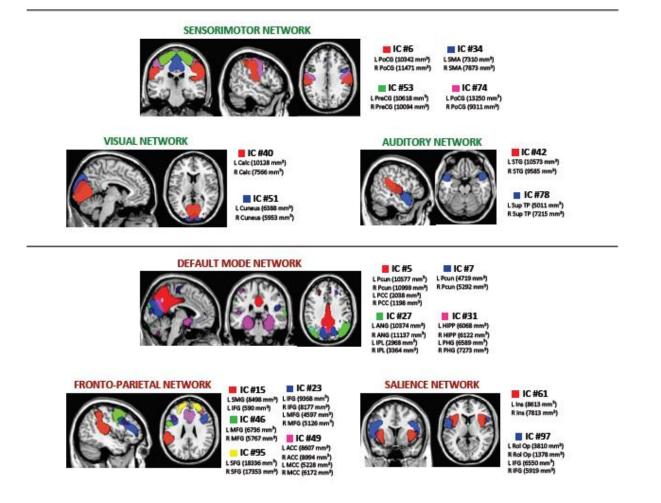


Figure e-1. Composite map of the relevant independent components (IC) obtained after the selection procedure. After running source-based morphometry with n=98 components, 26 relevant IC were selected and sorted into eight subcategories: cerebellar (4 components), subcortical (3), sensorimotor (4), visual (2), auditory (2), default-mode (5), fronto-parietal (5), and salience (4) networks. Each color in the composite map corresponds to a different IC within a given subcategory; the first IC of the network (in ascending order) is represented in red, the second one in blue, the third one in green, the fourth in violet and the fifth in red, as appropriate. IC patterns were thresholded at Z>2.5; spatial location and volume of the main clusters, having an extent>100 mm³, are reported. Images are in neurological convention. Abbreviations: L=left; R=right; Cer=cerebellum; Thal=thalamus; Putam=putamen; PoCG=postcentral gyrus; PreCG=precentral gyrus; Calc=calcarine cortex; STG=superior temporal gyrus; Sup TP=superior temporal pole; Pcun=precuneus; PCC=posterior cingulate cortex; MCC=middle cingulate cortex; ACC=anterior cingulate cortex; ANG=angular gyrus; IPL=inferior parietal lobule; HIPP=hippocampus; PHG=parahippocampal gyrus; SMG=supramarginal gyrus; IFG=inferior frontal gyrus; MFG=middle frontal gyrus; SFG=superior frontal gyrus; Rol Op=rolandic operculum.

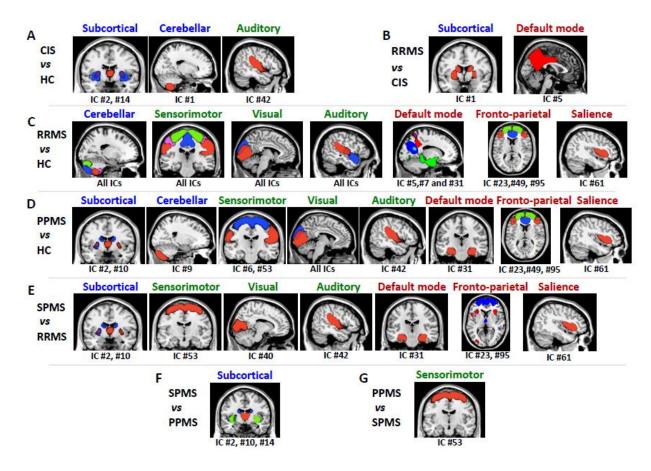


Figure e-2. Composite map of the relevant independent components (IC) showing significant baseline differences of grey matter atrophy between phenotypes at *post hoc* analysis. A) clinically isolated syndrome (CIS) patients *vs* healthy controls (HCs); B) CIS *vs* relapsing-remitting multiple sclerosis (RRMS) patients; C) RRMS patients *vs* HC; D) primary progressive multiple sclerosis (PPMS) patients *vs* HC; E) secondary progressive multiple sclerosis (SPMS) *vs* RRMS patients; F) SPMS *vs* PPMS patients; and G) PPMS *vs* SPMS patients. Each color in the composite map corresponds to a different IC within a given subcategory; the first IC of the network (in ascending order) is represented in red, the second one in blue, the third one in green, as appropriate. Images are in neurological convention.