White matter fiber-specific degeneration in older adults with metabolic syndrome

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ABSTRACT

Objective: Metabolic syndrome (MetS) is defined as a complex of interrelated risk factors for type 2 diabetes and cardiovascular disease, including glucose intolerance, abdominal obesity, hypertension, and dyslipidemia. Studies using diffusion tensor imaging (DTI) have reported white matter (WM) microstructural abnormalities in MetS. However, interpretation of DTI metrics is limited primarily due to the challenges of modeling complex WM structures. The present study used fixed-based analysis (FBA) to assess the effect of MetS on the fiber tract-specific WM microstructure in older adults and its relationship with MetS-related measurements and cognitive and locomotor functions to better understand the pathophysiology of MetS.

Methods: Fixed-based metrics, including microstructural fiber density (FD), macrostructural fiber-bundle cross-section (FC), and a combination of FD and FC (FDC), were evaluated in 16 healthy controls (no components of MetS; four men; mean age, 71.31 ± 5.06 years), 57 individuals with premetabolic syndrome (preMetS; one or two components of MetS; 29 men; mean age, 72.44 ± 5.82 years), and 46 individuals with MetS (three to five components of MetS; 27 men; mean age, 72.15 ± 4.97 years) using whole-brain exploratory FBA. Tract of interest (TOI) analysis was then performed using TractSeg across 14 selected WM tracts previously associated with MetS. The associations between fixed-based metrics and MetS-related measurements, neuropsychological, and locomotor function tests were also analyzed in individuals with preMetS and MetS combined. In addition, tensor-based metrics (i.e., fractional anisotropy [FA] and mean diffusivity [MD]) were compared among the groups using tract-based spatial statistics (TBSS) analysis.

Results: In whole-brain FBA, individuals with MetS showed significantly lower FD, FC, and FDC compared with healthy controls in WM areas, such as the splenium of the corpus callosum (CC), corticospinal tract (CST), middle cerebellar peduncle (MCP), and superior cerebellar peduncle (SCP). Meanwhile, in fixed-based TOI, significantly reduced FD was observed in individuals with preMetS and MetS in the anterior thalamic radiation, CST, SCP, and splenium of the CC compared with healthy controls, with relatively greater effect sizes observed in individuals with MetS. Compared with healthy controls, significantly reduced FD and FDC were only demonstrated in individuals in MetS, including regions with loss of FD, inferior cerebellar peduncle, inferior fronto-occipital fasciculus, MCP, and superior longitudinal fasciculus part I. Furthermore, negative correlations were observed between FD and Brinkman index of cigarette consumption cumulative amount and between FC or FDC and the Trail Making Test (parts B–A), which is a measure of executive function, waist circumference, or low-density lipoprotein cholesterol. Finally, TBSS analysis revealed that FA and MD were not significantly different among all groups.

Conclusions: The FBA results demonstrate that substantial axonal loss and atrophy in individuals with MetS and early axonal loss without fiber-bundle morphological changes in those with preMetS within the WM tracts are crucial to cognitive and motor function. FBA also clarified the association between executive dysfunction, abdominal obesity, hyper-low-density lipoprotein cholesterolemia, smoking habit, and compromised WM neural tissue microstructure in MetS.

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Keywords Diffusion MRI; Executive-motor dysfunction; Fixed-based analysis; Metabolic syndrome; Neurodegeneration; White matter

1. INTRODUCTION

Metabolic syndrome (MetS) is a complex of interrelated risk factors for type 2 diabetes and cardiovascular disease, including glucose intolerance, abdominal obesity, hypertension, and dyslipidemia, that is associated with increased mortality and morbidity rates [1]. It is a common public health and major clinical problem due to its continuing growth in prevalence worldwide [1]. The effects of MetS are multiple and wide-ranging. Recent studies have suggested the presence of widespread white matter (WM) microstructural abnormalities in
individuals with MetS, which lead to cognitive impairment and increased risk of dementia later in life [2]. In addition to cognitive dysfunctions, a poorer locomotor function has recently been observed in individuals with MetS [3]. Despite a clear connection, the pathological mechanisms underlying cognitive and locomotor impairments in MetS are not fully understood yet. Diffusion tensor imaging (DTI) is the current most widely used tool used to assess WM microstructural alterations. However, it is limited by its inability to represent the diffusion signal in crossing or kissing fiber regions. It has been suggested that approximately 90% of WM voxels at the current limit of resolution of diffusion-weighted imaging (2.4 mm isotropic voxel size) contain multiple fiber populations [4]. Furthermore, the tensor model has been shown to lack specificity regarding pathology [5]. Lower fractional anisotropy (FA) and/or higher mean diffusivity (MD) have been observed in the WM of individuals with MetS [6,7], indicating alterations in axonal diameter/density and/or myelin content [5]. Taken together, these limitations lead to an unreliable interpretation of DTI-associated metrics. Pixel-based analysis (FBA) was proposed to quantify the properties of WM fiber in its complex fiber geometry. The term “pixel” refers to a specific individual fiber population within a voxel [8,9] and is derived from WM fiber orientation distributions (FODs) as computed by constrained spherical deconvolution (CSD) method [10]. FBA enables the measurement of microstructural differences in fiber density (FD), macrostructural differences in the fiber-bundle cross-section (FC), or differences arising from the combination of FD and FC (fiber density and cross-section [FDC]) [9]. Adoption of the FBA framework has increased over the years, as summarized by Dholander et al. [11]. For example, studies using FBA in healthy adults reported widespread age-related decreases in FD, FC, and FDC indicative of WM degradation with age [12,13]. Changes in pixel-based metrics have also been observed in neurodegenerative diseases, such as Alzheimer’s disease [14] and Parkinson’s disease [15–18].

The present study used FBA to assess the effect of MetS on the WM fiber-specific microstructure in older adults and its relationship with MetS-related measurements and cognitive and locomotor functions to better understand the pathophysiology of MetS. Time-course gene expression profiles of a mouse model of MetS revealed a predisease state prior to MetS [19]. Therefore, we evaluated individuals with components of MetS who did not meet the criteria for diagnosis of MetS or premetabolic syndrome (preMetS) to determine the early signs of fiber tract-specific degeneration, which is considered the optimal time for preventive treatment [19]. Finally, we assessed DTI metrics for comparison.

2. PARTICIPANTS AND METHODS

2.1. Metabolic syndrome criteria

We defined MetS according to the guidelines of the International Diabetes Federation and the American Heart Association/National Heart, Lung, and Blood Institute [1]—which have been reported to be the most suitable for clinical practice and are ethnically specific [20]—as the presence of at least three of the following five risk factors: (1) abdominal obesity (waist circumference ≥ 85 cm for men and ≥ 90 cm for women, adjusted for Japanese individuals [21]); (2) elevated blood pressure (systolic pressure ≥ 130 mmHg and/or diastolic pressure ≥ 85 mmHg or receiving treatment for hypertension); (3) elevated fasting plasma glucose levels (> 100 mg/dL or receiving treatment for hyperglycemia); (4) elevated triglyceride levels (> 150 mg/dL or receiving treatment for hypertriglyceridemia); and (5) low levels of high-density lipoprotein (HDL) cholesterol (< 40 mg/dL or receiving treatment). Individuals with no risk factors were categorized as healthy controls, whereas those with one or two risk factors were categorized as preMetS [22].

2.2. Study participants

Data for the study participants were obtained from the Bunkyo Health Study [23], which was a prospective cohort study of older individuals (age, 65–84 years) in an urban community. The study protocol was approved by the ethics committee of Juntendo University, and all individuals provided written informed consent prior to participating in the study. Inclusion criteria were as follows: available and complete data for MetS-related characteristics, neuropsychological measures, locomotor function tests, assessment of WM hyperintensities (Fazekas scale), and brain 3T magnetic resonance imaging (MRI) results. Exclusion criteria were as follows: history of drug and/or alcohol abuse; past or present neurological or psychiatric disorders; dementia (mini-mental state examination [MMSE] score of ≤ 23) [24], depression (the Japanese version of the 15-item Geriatric Depression scale [GDS-15-J] score of ≥ 10) [25]; or structural brain MRI abnormalities.

Abbreviations

| Abbreviation | Description |
|--------------|-------------|
| AF | arcuate fasciculus |
| ATR | anterior thalamic radiation |
| CC | corpus callosum |
| CSD | constrained spherical deconvolution |
| CST | corticospinal tract |
| DTI | diffusion tensor imaging |
| FA | fractional anisotropy |
| FBA | flexible-based analysis |
| FC | fiber-bundle cross-section |
| FD | fiber density |
| FDC | fiber density and cross-section |
| FDR | false discovery rate |
| FODs | fiber orientation distributions |
| GDS-15-J | Japanese version of the 15-item Geriatric Depression Scale |
| HDL | high-density lipoprotein |
| HOMA-IR | homeostasis model assessment of insulin |
| ICP | inferior cerebellar peduncle |
| ICV | intracranial volume |
| IFO | inferior fronto-occipital fasciculus |
| ILF | inferior longitudinal fasciculus |
| LDL | low-density lipoprotein |
| MCP | middle cerebellar peduncle |
| MD | mean diffusivity |
| MetS | metabolic syndrome |
| preMetS | premetabolic syndrome |
| SCP | superior cerebellar peduncle |
| SLF | superior longitudinal fasciculus |
| TBSS | tract-based spatial statistics |
| TMT | trail making test |
| TOI | tract of interest |
| UF | uncinate fasciculus |
| WM | white matter |
2.3. Metabolic syndrome-related assessments

Among Japanese individuals, the cutoff values for waist circumference used in this study were based on the cutoff values of visceral fat area determined using computed tomography (CT) scan [26]. Visceral fat deposition at the navel level of >100 cm², corresponding to a waist circumference of 85 or 90 cm for Japanese males or females, respectively, is associated with higher prevalence of MetS risk factors. The cutoff value is larger for females because of the amount of subcutaneous fat is greater in women than in men when visceral adiposity is equivalent, although there is substantial variation between individuals. The areas of visceral and subcutaneous fat were measured using a 0.3 T MRI scanner after overnight fasting as described previously by Someya et al. [23]. Notably, the MRI measurements were in good agreement with the CT findings, which is considered the gold standard [27]. Arteriosclerosis was predicted using the cardio-ankle vascular index (CAVI), which is an index of overall arterial stiffness and is independent of blood pressure [28]. Insulin resistance was assessed using the homeostasis model assessment of insulin resistance (HOMA-IR) based on the fasting glucose and insulin plasma levels [29].

2.4. Neuropsychological tests

The present study evaluated global cognitive function, as measured by the Japanese version of the Montreal Cognitive Assessment (MOCA-J) [30] and MMSE [24], and the difference between time (in seconds) spent in completing parts B and A of the trail-making test (TMT [B–A]) to measure cognitive flexibility and executive function [31].

2.5. Locomotor function tests

The term “locomotive syndrome” was proposed by the Japanese Orthopaedic Association in 2007 to describe a condition of reduced mobility (sit-to-stand or gait) due to age-related impairment of the locomotor system, which includes the bone, cartilage, muscle, and nervous system [32]. The risk of locomotive syndrome is measured using the stand-up test, two-step test, and 25-question geriatric locomotive function scale (GLFS-25). Detailed descriptions of these tests are available in previous studies [33,34]. The two-step test score (length of two steps in cm) is standardized using the individual’s height (cm). The study participants were classified as no locomotive syndrome (stage 0), locomotive syndrome stage 1, or locomotive syndrome stage 2. Locomotive syndrome stage 1 (beginning to decline in mobility) is defined as a two-step test score of <1.3 and difficulty with one-leg standing from a 40-cm seat in the stand-up test (either leg), or a GLFS-25 score of ≥7 [33]. Locomotive syndrome stage 2 (progressing to a decline in mobility) was defined as a two-step test score of <1.1 and difficulty standing from a 20-cm seat using both legs in the stand test or a GLFS-25 score of ≥16 [33].

2.6. Imaging protocols

All imaging data were acquired using the 3 T system (MAGNETOM Prisma; Siemens Healthcare, Erlangen, Germany) with a 64-channel head coil. Three-dimensional T1-weighted images were obtained using a magnetization-prepared rapid gradient echo protocol (echo time [TE], 2.32 ms; repetition time [TR], 2300 ms; inversion time, 900 ms; field of view [FOV], 240 mm × 240 mm; matrix size, 256 × 256; slice thickness, 0.9 mm; and acquisition time, 5.21 min). Diffusion-weighted images were obtained using echo planar imaging in the anterior—posterior phase-encoding direction with the following parameters: TE, 70 ms; TR, 3300 ms; FOV, 229 mm × 229 mm; matrix size, 130 × 130; slice thickness, 1.8 mm; b-values, 1000 and 2000 s/mm²; 64 isotropic gradient directions; one non-diffusion-weighted (b = 0) volume; and acquisition time, 7.29 min. The standard and reverse phase-encoded blipped images with no diffusion weighting (blip up and blip down) were also obtained to correct for magnetic susceptibility-induced distortions related to echo-planar imaging acquisition [35].

2.7. Fixel-based analysis pipeline

FBA was performed according to the recommended pipeline [9,11]. Preprocessing steps of diffusion-weighted images—including denoising [36], removal of Gibbs ringing artifacts, eddy-current and motion-induced distortion correction [37], bias field correction [38], and up-sampling diffusion-weighted image spatial resolution to an isotropic voxel size of 1 mm [39]—were performed using MRtrix3 (http://mrtrix.org). We estimated FODs for each participant using the multishell multitissue-CSD algorithm with the group-average response functions of WM, gray matter, and cerebrospinal fluid [10]. The remaining processing steps included overall image intensity normalization of FOD images to enable the comparison of FOD amplitudes among participants using the median b = 0 intensity and the generation of a study-specific FOD template using FOD images from all participants with linear and nonlinear registration [39,40]. The study-specific population template was then transformed into MN152 common space using FA-based affine registration in the FMRIB Software Library 5.0.9 (FSL; Oxford Centre for Functional MRI of the Brain, UK; www.fmrib.ox.ac.uk/fsl). The following fixel-based metrics were calculated (Figure 1): (1) FD, which measures the intra-axonal volume of axons of a fiber population in each voxel compartment [9,11], (2) FC, which is a morphological measure of fiber-bundle cross-sectional size that is calculated as the extent of distortion in a bundle cross-section required to warp a participant’s FOD into the template image [9,11]. Before statistical analysis, FC was log-transformed (log-FC) to ensure zero-centered and normally distributed data. (3) FDC, which is a combined measure of FD and FC (multiplicative) that accounts for both microscopic and macroscopic changes in a fiber bundle, thereby providing sensitivity to any differences related to the capacity of the WM to transmit information [9,11].

2.8. Tract of interest analysis

Tract of interest (TOI) analysis was performed to investigate further potential WM degeneration of fiber pathways in individuals with pre-MetS and MetS. WM tract segmentation was conducted using the TractSeg algorithm [41]. TractSeg was applied to the study-specific population template, which had been transformed into MN152 common space [42]. We delineated WM tracts in Figure 2 previously implicated in MetS or individual components of MetS including arcuate fasciculus (AF), anterior thalamic radiation (ATR), cingulum, corticospinal tract (CST), inferior cerebellar peduncle (ICP), inferior fronto-occipital fasciculus (IFO), inferior longitudinal fasciculus (ILF), middle cerebellar peduncle (MCP), superior cerebellar peduncle (SCP), splenium of the corpus callosum (CC), superior longitudinal fasciculus (SLF-I), SLF-II, and SLF-III, and uncinate fasciculus (UF) [2]. The anatomical definition and functions of the WM regions are described in Supplementary Table 1. The FD, log-FC, and FDC maps from each participant were then cropped to only include fixels belonging to the tracts of interest, leaving 14 (splenium of the CC and bilateral AF, ATR, cingulum, CST, ICP, IFO, ILF, MCP, SCP, SLF-I, SLF-II, SLF-III, and UF) fixel maps for each fixel-based metric. These fixel maps were then statistically analyzed.
2.9. Tensor-derived metrics

The DTIFIT tool, which is a part of the FSL, was used to generate FA and MD maps for each participant using preprocessed diffusion-weighted data with b-values of 0 and 1000 s/mm². Between-group comparisons of FA and MD were performed using tract-based spatial statistics (TBSS) [43], which is currently the most commonly used voxel-wise...
analyses were displayed using the TBSS procedure was as follows: (1) FA maps of all participants were aligned in 1 × 1 × 1 mm³ MNI152 common space (an averaged brain) using the nonlinear registration tool in the FMRIB Software Library 5.0.9 (Oxford Centre for Functional MRI of the Brain, UK; www.fmrib.ox.ac.uk/fsl) [44]. (2) A population-based mean FA image was created and thinned to generate a mean FA skeleton representing the centers of all tracts common to the group. The threshold FA level of this skeleton was 0.2 to exclude voxels from gray matter and cerebrospinal fluid. (3) The averaged FA map for each participant was projected onto the skeleton. Finally, the mean diffusivity maps were projected onto the FA-derived skeleton after applying each participant’s warping registration to the common space.

2.10. Intracranial volume measurements
The estimated intracranial volume (ICV) was obtained for each study participant using T1-weighted images in FreeSurfer version 6.0.0 (http://surfer.nmr.mgh.harvard.edu/fswiki) with the recon-all pipeline [45]. The ICV was used as a nuisance covariate in statistical analyses.

2.11. Statistical analysis
Kolmogorov–Smirnov test was used to assess the normality of the data in IBM SPSS Statistics version 27.0 (IBM Corporation, Armonk, NY, USA). Demographic and clinical data were analyzed using one-way analysis of variance with Tukey’s honest significant difference post hoc test or Kruskal–Wallis with Mann–Whitney U post hoc test for normally or nonnormally distributed continuous data, respectively, whereas chi-square test was used for categorical variables. P-values (two-tailed) of <0.05 were considered statistically significant. For each fixed-metric wise, a connectivity-based fixed enhancement for statistical inference [8] with default smoothing parameters (smoothing, 10 mm full-width at half maximum; C, 0.5; E, 2, and H, 3) was performed in whole-brain (including 2 million streamlines from the template tractogram) and TOI analyses. A general linear model framework was applied to compare: (1) healthy controls vs. preMetS; (2) healthy controls vs. MetS; and (3) preMetS vs. MetS. Age, sex, and years of education were included as nuisance covariates. Log-transformed ICV (log-ICV) was included as a nuisance covariate for log-FC and FDC (but not FD) to avoid false-positive results with global effects of brain scaling resulting from the registration to a template removed [9,46]. Family-wise error (FWE)-corrected P-values were then assigned to each fixed using nonparametric permutation testing with 10,000 permutations. FWE-corrected P-values <0.05 were considered statistically significant for whole-brain FBA. Statistical analyses were conducted separately for each tract in the TOI analysis; thus, FWE-corrected P-value at the peak fixed level within each tract were then submitted for false discovery rate (FDR) correction for multiple comparisons across 14 WM tracts. Tracts with peak pixel FWE-corrected P-values <0.05 following FDR correction were considered to withstand correction for multiple comparisons across tracts. Significant fixed in both whole-brain and TOI statistical analyses were displayed using the mrview tool in FMRIBx3. Streamlines from template-derived tractogram were cropped to include only fixed that were significant (FWE-corrected P < 0.05) and color-coded by the effect size expressed as a percentage relative to the control group. We also assessed the associations between mean values of fixed-based metrics along each tested WM tract (Supplementary Table 2) and the MetS-related measurements, neuropsychological tests, locomotor function tests (Table 1) in individuals with preMetS and MetS combined using partial correlation tests adjusting for age, sex, and years of education in SPSS 27. FDR-adjusted (for 14 WM tracts) P-values of <0.05 (two-tailed) were considered significant.

For TBSS analysis, a voxel-wise general linear model framework, including age, sex, years of education, and log-ICV as covariates, was used to compare FA and MD between all groups using the FSL randomize tool with 10,000 permutations. Between-group differences were considered significant at P < 0.05 and corrected for multiple comparisons using the FWE and threshold-free cluster enhancement approaches.

3. RESULTS

3.1. Demographic and clinical assessments
A total of 119 older adults were enrolled in the present study: 16 healthy controls (4 men and 12 women; mean age, 71.31 ± 5.06 years), 57 individuals with preMetS (29 men and 28 women; mean age, 72.44 ± 5.82 years), and 46 individuals with MetS (27 men and 19 women; mean age, 72.15 ± 4.97 years). There were no differences in age; sex; years of education; Brinkman index of cigarette consumption cumulative amount; handedness; total cholesterol; low-density lipoprotein (LDL) cholesterol; mean CAVI; Fazekas scale (ventricular and deep WM); locomotive syndrome stage; or the MOCA-J, MMS, TMT (B–A), GDS-15-J, and GLFS-25 scores among the groups. No significant differences were observed between healthy controls and individuals with preMetS for HDL cholesterol, triglycerides, fasting plasma glucose, HbA1c, HOMA-IR, and standardized two-step test score, or between individuals with preMetS and MetS for systolic and diastolic blood pressures and stand-up test score. Individuals with preMetS and MetS showed significantly (P < 0.05) higher values for body mass index, waist circumference, systolic and diastolic blood pressures, subcutaneous and visceral fat areas, and lower stand-up test scores than healthy controls. Individuals with MetS showed significantly lower HDL cholesterol and standardized two-step test scores, and higher triglycerides, fasting plasma glucose, HbA1c, and HOMA-IR than healthy controls. Furthermore, individuals with MetS had significantly lower HDL cholesterol and higher body mass index, waist circumference, triglyceride level, fasting plasma glucose, HbA1c, HOMA-IR, subcutaneous and visceral fat areas, and two-step test scores than those with preMetS.

3.2. Whole-brain fixed-based analysis
Compared with healthy controls, individuals with MetS had significantly (FWE-corrected P < 0.05) lower FD in the splenium of the CC and right posterior limb of the internal capsule, lower log-FC in the left CST, bilateral mammillothalamic tract, cerebral peduncle, and SCP, and lower FDC in the bilateral CST, mammillothalamic tract, MCP, and SCP (Figure 3). No significant differences were observed in any of the fixed-based metrics between healthy controls and individuals with preMetS or between individuals with preMetS and MetS.

3.3. Fixed-wise tract of interest analysis
Individuals with MetS and preMetS showed significantly (FWE-corrected P < 0.05 following FDR correction) reduced FD (Figure 4) in the ATR (left), CST (bilateral), SCP (bilateral), and splenium of the CC than healthy controls, and the differences were relatively more pronounced in those with MetS. Individuals with MetS showed significantly reduced log-FC (Figure 5) in the ATR (left), CST (bilateral), ICP (bilateral), MCP (bilateral), SCP (bilateral), and SLF-I (left) compared with healthy controls. Individuals with MetS also showed a significantly reduced FDC (Figure 6) in the ATR (left), CST (bilateral), ICP (left), IFO (left), MCP, and SCP (bilateral) compared with healthy controls. No significant differences were observed for FDC or log-FC between healthy controls and
Table 1 — Demographics and clinical characteristics of the participants.

|                          | HCs (N = 16) | preMetS (N = 57) | MetS (N = 46) | P-values |
|--------------------------|--------------|------------------|---------------|---------|
|                          |              |                  |               | HC vs. preMetS vs. MetS | HC vs. MetS | HC vs. preMetS | MetS vs. preMetS | p-values |
| Sex, n (male/female)     |              |                  |               | 0.067   |
| Age (years)              |              |                  |               | 0.841   |
| Years of education       |              |                  |               | 0.355   |
| Brinkman index           |              |                  |               | 0.072   |
| Handedness, n (left/right/mixed) |           0/14/2   | 3/51/3           | 9/45/3        | 0.197   |
| MetS-related characteristics: |          |                  |               |         |
| BMI (kg/m²)              |              |                  |               | <0.001  |
| Waist circumference (cm) |              |                  |               | <0.001  |
| Systolic blood pressure (mmHg) |           |                  |               | <0.001  |
| Diastolic blood pressure (mmHg) |           |                  |               | <0.001  |
| Total cholesterol (mg/dL) |              |                  |               | 0.374   |
| HDL cholesterol (mg/dL)  |              |                  |               | 0.008   |
| LDL cholesterol (mg/dL)  |              |                  |               | 0.122   |
| Triglycerides (mg/dL)    |              |                  |               | <0.001  |
| Fasting plasma glucose (mg/dL) |           |                  |               | <0.001  |
| HbA1c (mmol/mol)         |              |                  |               | 0.005   |
| HOME-IR index           |              |                  |               | 0.002   |
| Visceral fat area (cm²)  |              |                  |               | <0.001  |
| Subcutaneous fat area (cm²) |            |                  |               | <0.001  |
| Arteriosclerosis measure: |          |                  |               |         |
| Bilateral CAV⁰          |              |                  |               | 0.126   |
| Neuropsychological measures: |         |                  |               |         |
| MOCA-J⁰                 |              |                  |               | 0.669   |
| MMSE⁰                   |              |                  |               | 0.542   |
| TMT (B–A; seconds)⁰     |              |                  |               | 0.790   |
| GDS-15-J                |              |                  |               | 0.238   |
| Locomotor function tests: |          |                  |               |         |
| Stage, N (0/1/2/3)⁰     |              |                  |               | 0.149   |
| Stand-up test³           |              |                  |               | 0.004   |
| Standardized two-step test³ |           |                  |               | 0.006   |
| GLF-S 25⁰               |              |                  |               | 0.722   |
| Fazekas scale:           |              |                  |               |         |
| Perventricular WM, N (0/1/2/3)⁰ |     0/14/2/0   | 0/46/11/0        | 1/39/6/0     | 0.655   |
| Deep WM, N (0/1/2/3)⁰   |              |                  |               | 0.792   |

Data are presented in mean ± standard deviation unless otherwise stated. Statistical analyses were performed using χ² test, one-way analysis of variance with Tukey’s Honest Significant Difference post hoc test, and Kruskal–Wallis with post hoc Mann–Whitney test. Bold values denote statistical significance (P < 0.05).

MOCA-J, Montreal Cognitive Assessment; MMSE, Mini-Mental State Examination; TMT, Trail Making Test; GDS-15-J, Japanese version of the 15-item Geriatric Depression Scale; GLF-S 25, 25-question geriatric locomotive function scale; HOME-IR, homeostasis model assessment of insulin resistance; LDL, low-density lipoprotein; MetS, metabolic syndrome; MOCA-J, Japanese version of Montreal cognitive assessment; MMSE, mini-mental state examination; preMetS, premetabolic syndrome; TMT (B–A), the difference between time in seconds spent in completing parts B and A of the trail-making test; WM, white matter.

Figure 3: Fiber tract-specific reductions in individuals with metabolic syndrome compared to healthy controls from whole-brain fixed-based analysis. Streamline segments (blue) were cropped from the template tractogram to include only streamline points corresponding to significant fixels (Family-wise error-corrected P < 0.05) and colored according to the percentage effect decrease in patients with metabolic syndrome compared with healthy controls. FD, fiber density; FDC, fiber density and cross-section; log-FC, log-transformed fiber-bundle cross-section.
individuals with preMetS or any of the pixel-based metrics between individuals with preMetS and MetS. We observed a significant (FDR-adjusted $P < 0.05$) correlation between increased Brinkman index and reduced FD in the MCP and splenium of the CC (Table 2). Significant negative correlations were also observed between log-FC and FDC metrics and waist circumference, LDL cholesterol, and TMT (BeA) scores. Specifically, associations were demonstrated between increased waist circumference and reduced log-FC in the AF, ATR, cingulum, CST, IFO, ILF, SLF-I, SLF-II, SLF-III, and UF or lower FDC in the AF, ATR, cingulum, ICP, MCP, SCP, and UF or FDC in the ICP, ILF, and UF; and between higher TMT (B–A) scores and lower log-FC in the AF, ATR, cingulum, CST, ICP, IFO, ILF, MCP, SCP, SLF-I, SLF-II, SLF-III, and UF or lower FDC in the ATR, cingulum, CST, ICP, MCP, and SCP.

3.4. Tract-based spatial statistics analysis

There were no significant differences in FA and MD among all groups.

4. DISCUSSION

In the present study, we applied FBA to investigate WM fiber tract-specific in older adults with preMetS and MetS. Individuals with MetS exhibit extensive axonal loss and fiber-bundle atrophy within WM pathways considered critical to cognitive and motor functions, which are largely congruent with the regions of WM abnormalities previously reported in MetS [7] or individual components of MetS [2]. Meanwhile, individuals with preMetS showed subtle axonal loss without evidence of morphological alterations to the fiber bundle. Our findings also showed evidence of associations between MetS-related WM neuronal damage and impairment in executive function, abdominal obesity, hyper-LDL cholesterolemia, or smoking habit.

4.1. Reduced white matter fiber tract-specific in premetabolic syndrome and metabolic syndrome

Although there is limited histological evidence, particularly in WM, lower neuronal cell bodies have been observed in various frontal and temporal areas in the postmortem brain tissue of donors who were obese, a main feature of MetS, compared with normal-weight individuals [47]. In agreement, pixel-wise changes in individuals with preMetS and MetS were identified in the WM tracts that connect parts of the frontal or temporal cortices to other cortical or subcortical areas, such as the ATR, CST, IFO, and SLF-I. Thus, the possibility that the distribution of WM alterations among individuals with preMetS and MetS is a primary effect of WM pathology and/or axonal degeneration associated with cortical neuron injury warrants further investigation.
Our observations in individuals with preMetS and MetS indicated substantial axonal loss (i.e. reduced FD) in the ATR, CST, SCP, and splenium of the CC compared with healthy controls, with relatively greater effect sizes observed in individuals with MetS. FBA also enabled the identification of fiber-bundle atrophy (i.e., reduced log-FC) only in individuals with MetS. Notably, fiber tract-specific atrophy was

Figure 5: White matter tracts showing reduced log-transformed fiber-bundle cross-section (log-FC) in individuals with metabolic syndrome (MetS) compared with healthy controls (HCs). Tract-specific streamline segments (blue) represent streamlines corresponding to significant pixels (Family-wise error [FWE]-corrected $P < 0.05$) and are colored according to the percentage effect decrease in individuals with MetS compared with HCs. Boxplots illustrate within-group variability (calculated across all significant pixels). The bottom and top of the box represent the first and third quartiles, respectively, and the band inside the box represents the median. Whiskers represent the maximum and minimum values of all data. FWE-corrected $P$-values are shown at the peak pixel level of each tract after false discovery rate correction. ATR, anterior thalamic radiation; CST, corticospinal tract; ICP, inferior cerebellar peduncle; MCP, middle cerebellar peduncle; SCP, superior cerebellar peduncle; SLF, superior longitudinal fasciculus.

Figure 6: White matter tracts showing reduced fiber density and cross-section (FDC) in individuals with metabolic syndrome (MetS) compared with healthy controls (HCs). Tract-specific streamline segments (blue) represent streamlines corresponding to significant pixels (Family-wise error [FWE]-corrected $P < 0.05$) and are colored according to the percentage effect decrease in individuals with MetS compared with HCs. Boxplots illustrate within-group variability (calculated across all significant pixels). The bottom and top of the box represent the first and third quartiles, respectively, and the band inside the box represents the median. Whiskers represent maximum and minimum values of all data. FWE-corrected $P$-values are shown at the peak pixel level of each tract after false discovery rate correction. ATR, anterior thalamic radiation; CST, corticospinal tract; ICP, inferior cerebellar peduncle; IFO, inferior fronto-occipital fasciculus; MCP, middle cerebellar peduncle; SCP, superior cerebellar peduncle.
observed spatially more extensive, including in regions with loss of FD, ICP, MCP, and SLF-I. As expected, individuals with MetS showed reduced FDC overlap with regions with significantly reduced FD and FC, with relatively larger effect sizes than the changes in FD or log-FC alone. FDC combines the information of microstructural differences in FD and morphological changes in FC, thus enabling a more sensitive assessment of fiber-wise effects [11].

The dissociable results of fiber-based metrics in individuals with preMetS (only reduced FD) and MetS (reduced in all fiber metrics) indicated that different WM tracts are affected differently or show different stages of neurodegeneration. FD provides a measure approximately proportional to axons present within a voxel and depends on axons as well as the nonaxonal space or volume in between [11]. It was suggested that the number of axons or their individual volume within the fiber-specific tract may be reduced in the earlier stages of degeneration, but the macroscopic FC remains unchanged if extracellular area and matrix related to gliosis or inflammation filled the additional axonal space. The fiber bundle becomes atrophic after the debris is cleared [9,48], implying that axonal loss could precede fiber bundle atrophy. A longitudinal study of Parkinson’s disease, for example, showed reductions in FD between the first follow-up assessment and baseline, whereas FC only decreased between the second and first follow-up assessment [17]. Our findings also suggest that log-FC and FDC were reduced without changes in FD in individuals with MetS compared with healthy controls, noticeably in the ICP and MCP. In some cases, if degeneration is followed by atrophy (i.e., reduced FC), the within-voxel FD for the remaining axons may suggest an increase in FD, resulting in insignificant results in between-group comparisons [9]. Nevertheless, since reduced FDC, which is a combined measure of FD and FC, was also observed, it is reasonable to assume that the changes in ICP and MCP were driven by both fiber density and bundle cross-sectional size. Collectively, our observations indicate early axonal loss in preMetS and extensive axonal loss as well as fiber-bundle atrophy in MetS. In support of our findings, a previous longitudinal study showed subclinical elevation in MetS risk indicators as predictors of greater WM microstructural damage [49]. Furthermore, axonal atrophy was associated with irreversible disability [48]; thus, detecting axonal loss before atrophy may be necessary for early intervention.

In terms of laterality, changes in pixel metrics were primarily observed in the left hemisphere in some WM tracts. While we have no clear explanation for these observations, we speculate that disorders of the brain vascular system may be related to our findings. Hypertension is known to contribute to vascular damage and plaque formation in cerebral blood vessels [50]. The direct connection of the left carotid artery to the aortic arch has been shown to cause higher arterial pressure, which leads to greater vulnerability of atherosclerotic disease in the left carotid artery [51]. In a recent study involving elderly individuals with hypertension, abnormalities in the perivascular space diffusion signal, indicating brain clearance dysfunction, were observed in the left hemisphere only [52]. Nevertheless, some changes were also observed bilaterally, indicating that the pathological processes in MetS are complex and involve multiple mechanisms that need further exploration.

### 4.2. Effects of metabolic syndrome on executive and locomotor functions

Our results revealed significant associations between fiber density and bundle cross-sectional reductions and executive dysfunction in individuals with preMetS and MetS, as indicated by prolonged time to complete TMT (B–A), in WM tracts important for executive function [53], such as ATR, cingulum, IFO, ILF, and SLF. The WM microstructural damage, such as lower FA and/or higher MD, has also been associated with impairment in executive function in patients with individual components of MetS, such as hyperglycemia [54], hypertension [55], and obesity [56]. However, in contrast to DTI, FBA may provide more detailed biological information on changes occurring in the WM structure related to executive dysfunction in MetS. It is worth noting that we found no significant difference in TMT (B–A) scores between groups, which could suggest that the WM changes gradually and begins long before symptoms emerge. Accordingly, our findings indicated fiber-based metrics as promising early biomarkers of executive dysfunction in MetS.

Furthermore, we observed poorer locomotor function in individuals with MetS compared with healthy controls and individuals with
Finally, we noted an inverse relationship between FD and Brinkman and risk factor for vascular-based disease. Thus, LDL cholesterol cholesterol and is considered to be the major atherogenic lipoprotein [70]. LDL cholesterol represents 60% of lipid measurements (HDL cholesterol, triglycerides, and total cholesterol) [70].

4.3. Modifiable lifestyle variables associated with white matter neural tissue microstructure in metabolic syndrome

Higher waist circumference was associated with reduced log-FC or FDC for approximately all WM tracts in individuals with preMetS and MetS. However, unexpectedly, Birdsell et al. [63] found that higher FA was significantly associated with higher waist circumference. DTI cannot resolve crossing fibers accurately; thus, higher FA may reflect selective neurodegeneration or compensatory WM microstructural reorganization [15]. Alternatively, the CSD model can accurately resolve different fiber orientations within a voxel of interest [11], providing a more reliable measurement. Waist circumference is a common indicator of abdominal fat accumulation [64], which is a major feature of MetS. The Japanese waist circumference threshold is influenced by both visceral and subcutaneous fat [26]. A large study involving Japanese participants revealed that visceral and subcutaneous adipose tissue accumulation was associated with insulin resistance [65]. Abdominal obesity-induced neuroinflammation and brain insulin resistance are associated with breakdown of the blood–brain barrier [66] and reduced intraneuronal glucose metabolism, respectively, in MetS [67], and have been suggested as the pathological features underlying WM neurodegeneration. However, we did not find any evidence of a direct association between insulin resistance, as indexed by HOMA-IR, and WM microstructure. Higher HOMA-IR was correlated with lower brain glucose metabolism, as observed using fluorodeoxyglucose positron emission tomography; however, this association has only been detected in amyloid-positive patients [68]. Accordingly, future studies should use fix-based measures to clarify the relationship between MetS and Alzheimer’s disease.

We observed that higher levels of LDL cholesterol negatively impact WM microstructure, as indicated by lower log-FC or FDC. Our findings were consistent with those from previous studies that evaluated associations between serum lipid levels and DTI measurements in the WM of healthy adults [69,70]. For instance, increased levels of LDL cholesterol were correlated with reduced FA and LDL cholesterol had the largest number of voxels, showing significant associations with FA, compared to other lipid measurements (HDL cholesterol, triglycerides, and total cholesterol) [70]. LDL cholesterol represents 60%–70% of circulating serum cholesterol and is considered to be the major atherogenic lipoprotein and risk factor for vascular-based disease. Thus, LDL cholesterol potentially influenced WM microstructure by promoting intracranial atherosclerosis and brain chronic hyperperfusion [71].

Finally, we noted an inverse relationship between FD and Brinkman index, which reflects cumulative cigarette consumption, in the splenium of CC. Our findings are concordant with those of previous studies that demonstrated a significant correlation between reduced FA in the CC and tobacco consumption [72] and duration of cigarette smoking [73]. A previous study showed that smokers were relatively more insulin-resistant than nonsmokers and that smoking mediates endothelial dysfunction associated with reduced arterial compliance and increases arterial stiffness [74].

4.4. Advantages of fix-based over tensor-derived metrics

To date, scarce data are available regarding the effects of MetS on WM neural tissue microstructure using DTI, particularly in older adults with MetS, and the findings of previous studies are inconsistent. Segura et al. [6] observed compromised WM microstructure (as indexed by lower FA and higher MD) occurring predominantly in the anterior parts of the CC, ILF, SLF, and UF in individuals with MetS compared with healthy controls. In contrast, the results of the study by Sala et al. [75] and the present study showed no significant differences in FA and MD between individuals with preMetS or MetS and healthy controls. This discrepancy may be due to the differences in the criteria used to define MetS as well as study sample characteristics. The inconsistencies in DTI findings may also be attributable to the limitations of the tensor-derived metrics in regions with crossing fibers. In contrast to the findings of the study by Segura et al. [6], we did not observe differences in the anterior–posterior gradient using FBA. Indeed, Segura et al. reported that WM regions exhibit significant changes in DTI metrics, such as CC and SLF [6], and have been found to contain a relatively higher number of crossing fibers per voxel, suggesting unreliable interpretation of DTI-associated metrics [4]. In accordance with the study by Sala et al. [75], we were unable to demonstrate WM microstructural changes in individuals with preMetS using DTI. This may indicate that fix-based metrics have a higher sensitivity than DTI in detecting early axon changes, as was also shown previously in studies in neurodegenerative diseases, such as Alzheimer’s disease [14], Parkinson’s disease [15,18], and multiple sclerosis [48]. This fix-based approach allows us to differentiate between changes in microstructural FD and macrostructural FC; thus, provides greater sensitivity in detecting groupwise differences in preMetS, particularly in crossing fiber regions.

4.5. Study limitations and future directions

The present study has some limitations. Although we included more older adults with MetS than previous studies (Segura et al. [6], n = 19, and Sala et al. [75], n = 18), the present study had a small sample size (n = 46) in relation to the prevalence of MetS and may have limited statistical power. This might have led to false-positive or false-negative results. Furthermore, the cross-sectional study design of the study limited our ability to examine degenerative changes in WM over the progression of preMetS and MetS. Future studies with larger sample sizes investigating the longitudinal trajectory of individuals with preMetS who progress to MetS are likely to provide further insight into the association between clinical trajectory and degenerative changes in particular fiber pathways.

The prevalence of MetS varies and depends on the MetS criteria used [20] as well as the ethnic group. In the present study, we only included Japanese elderly living in urban areas and adjusted the waist circumference threshold for our Japanese cohort. Further studies are required to investigate changes in WM in MetS defined using different criteria in different ethnicities. Finally, MetS has been liked with a higher WM hyperintensity load [76]. Although the present study showed no statistical differences in peri-ventricular and deep WM Fazekas scales between all groups and correlation analysis with fix-based metrics, the presence of these
lesions may be associated with decreases in pixel-wise metrics [77]. Future studies are required to investigate the association of pixel-based metrics and WM lesions.

4.6. Conclusions
Our findings suggest that individuals with MetS have a characteristic pattern of WM fiber tract degeneration, involving WM pathways considered crucial for cognitive and motor functions, that manifests as both axonal loss and atrophy, whereas individuals with preMetS show earlier WM degenerative process that manifests as reduced FD without fiber-bundle morphology changes. Furthermore, we confirmed the negative impact of MetS on executive function performance associated with compromised WM neural tissue microstructure. Finally, our findings indicate that optimal management of modifiable variables, such as abdominal obesity, hyper-LDL-cholesterolemia, and smoking habits, are necessary to maintain WM neural health in older adults.

DATA STATEMENT

The data that support the findings of this study are available on request from the corresponding author.

AUTHOR CONTRIBUTIONS

Christina Andica: Conceptualization; Methodology; Formal analysis, Visualization, and Writing - original draft. Koji Kamagata: Conceptualization, Methodology; and Writing - review & editing, and Funding acquisition. Wataru Uchida, Kaito Takabayashi, and Keigo Shimoji: Formal analysis, Visualization, and Writing — review & editing. Hideyoshi Kaga, Yuki Someya, Yoshitumi Tamura, Ryuzu Kawamori, Hirotaka Watada, and Masaaki Hori: Conceptualization. Methodology, Supervision, Writing - review & editing. Shigeki Aoki: Conceptualization, Writing - review & editing, Supervision, and Funding acquisition.

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CONFLICT OF INTEREST

None.

APPENDIX A. SUPPLEMENTARY DATA

Supplementary data to this article can be found online at https://doi.org/10.1016/j.molmet.2022.101527.

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