A Metabolic Profile of Polyamines in Parkinson Disease: A Promising Biomarker

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Objective: Aging is the highest risk factor for Parkinson disease (PD). Under physiological conditions, spermidine and spermine experimentally enhance longevity via autophagy induction. Accordingly, we evaluated the ability of each polyamine metabolite to act as an age-related, diagnostic, and severity-associated PD biomarker.

Methods: Comprehensive metabolome analysis of plasma was performed in Cohort A (controls, n = 45; PD, n = 145), followed by analysis of 7 polyamine metabolites in Cohort B (controls, n = 49; PD, n = 186; progressive supranuclear palsy, n = 19; Alzheimer disease, n = 23). Furthermore, 20 patients with PD who were successively examined within Cohort B were studied using diffusion tensor imaging (DTI). Association of each polyamine metabolite with disease severity was assessed according to Hoehn and Yahr stage (H&Y) and Unified Parkinson’s Disease Rating Scale motor section (UPDRS-III). Additionally, the autophagy induction ability of each polyamine metabolite was examined in vitro in various cell lines.

Results: In Cohort A, N8-acetylspermidine and N-acetylputrescine levels were significantly and mildly elevated in PD, respectively. In Cohort B, spermine levels and spermine/spermidine ratio were significantly reduced in PD, concomitant with hyperacetylation. Furthermore, N1,N8-diacetylspermidine levels had the highest diagnostic value, and correlated with H&Y, UPDRS-III, and axonal degeneration quantified by DTI. The spermine/spermidine ratio in controls declined with age, but was consistently suppressed in PD. Among polyamine metabolites, spermine was the strongest autophagy inducer, especially in SH-SY5Y cells. No significant genetic variations in 5 genes encoding enzymes associated with spermine/spermidine metabolism were detected compared with controls.

Interpretation: Spermine synthesis and N1,N8-diacetylspermidine may respectively be useful diagnostic and severity-associated biomarkers for PD.

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Aging is the major risk factor for many chronic disorders, including diabetes mellitus, cancer, cardiovascular diseases, and neurodegenerative diseases including Parkinson disease (PD), whose current treatment is limited to symptomatic relief.1–3 Because PD prevalence increases with age, the number of PD patients is estimated to double from 6.9 million in 2015 to 14.2 million in 2040.4 Although age-related molecular mechanisms of PD (eg, dopamine metabolism, iron accumulation, mitochondrial DNA changes, and decreased protein-degradation efficiency) have been proposed,5 no corresponding blood biomarkers have been validated for widespread clinical use. Blood-based biomarkers associated with aging-related risk for PD would enable efficient monitoring of the disease process and could be used for development of therapies.

Polyamines are ubiquitous small polycations that ionically bind to various negatively charged molecules and have many functions, mostly linked to cell growth, survival, and proliferation.6 Prime examples of polyamines are putrescine, spermidine (Spd) and spermine (Spm), whose levels are strictly controlled. Putrescine is sequentially converted into Spd and Spm by S-(Spd) and spermine (Spm), whose levels are strictly controlled. Blood-based biomarkers associated with aging-related risk for PD would enable efficient monitoring of the disease process and could be used for development of therapies.

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Putrescine is sequentially converted into Spd and Spm by S-(Spd) and spermine (Spm), whose levels are strictly controlled. Blood-based biomarkers associated with aging-related risk for PD would enable efficient monitoring of the disease process and could be used for development of therapies.

FIGURE 1: Polyamine metabolism in humans. Schematic shows polyamines (putrescine, spermidine, and spermine), and their acetylated metabolites and converting enzymes. Ac-CoA = acetyl-coenzyme A; APAO = N1-acetyl polyamine oxidase; ATP = adenosine triphosphate; CoA = coenzyme A; MTA = 5′-deoxy-5′-methylthioadenosine; SAMHC = S-adenosylmethyl homocysteamine; SAT1/2 = spermidine/spermine acetyltransferase 1/2.

cSpm) are produced by acetylation of Spd, and their acetylated metabolites and converting enzymes. Ac-CoA = acetyl-coenzyme A; APAO = N1-acetyl polyamine oxidase; ATP = adenosine triphosphate; CoA = coenzyme A; MTA = 5′-deoxy-5′-methylthioadenosine; SAMHC = S-adenosylmethyl homocysteamine; SAT1/2 = spermidine/spermine acetyltransferase 1/2.

Subjects and Methods
Ethics Statement
This study protocol complied with the Declaration of Helsinki and was approved by the ethics committee of Juntendo University (#2012157). Written informed consent was obtained from all participants.

Participants
All participants were recruited at the Juntendo University Hospital and examined by board-certified neurologists. Cohort A (previously reported as “2nd cohort” in our report16) and Cohort B were independently recruited from December 2014 to February 2015 and from December 2016 to January 2017, respectively. PD, PSP, and AD were diagnosed according to the well-established criteria for each disease.17–19 We excluded PD patients with possible dementia (Mini-Mental State Examination score < 24) to avoid substantial overlap between PD with possible dementia and AD. Neither participants nor controls had a history of tumors, cancer, aspiration pneumonia, or inflammatory diseases including collagen vascular diseases. Participants suffering from acute infectious diseases or acute/chronic renal or hepatic failure at the time of sample collection were also excluded. Disease duration means the time since initial motor symptoms of PD. Hoehn and Yahr stages (H&Y) and...
Unified Parkinson’s Disease Rating Scale motor section (UPDRS-III) scores were defined during the “on phase” for practical and ethical reasons. UPDRS-III score in Cohort B may be at a lower level because of the preserved general physical status without any critical illness. L-dopa equivalent dose (LED) was calculated based on a previous report.20

Sample Collection
Sample collection was performed from December 2014 to February 2015 and from December 2016 to January 2017 for Cohorts A and B, respectively. Plasma and serum were extracted as described previously.16 Each 500μl of plasma or serum aliquot was stored in a −80°C freezer until use. Sample preparation followed by immediate mass spectrometry analysis of Cohorts A and B were performed in February 2015 and April 2017, respectively.

Sample Preparation
Polyamine metabolites were extracted as described previously.21 Briefly, 200μl of serum was immediately added to 200μl of 10% trichloroacetic acid containing 10pmol of N1-Ac-Spd-d6 (Toronto Research Chemical, Toronto, Ontario, Canada) as an internal standard. The solution was centrifuged at 15,000 rpm for 15 minutes at 4°C, and the upper aqueous layer was collected. The sample was dissolved in 50μl of Milli-Q water after lyophilizing, and 10μl was used for liquid chromatography—mass spectrometry (LC-MS). The concentration of each compound was estimated by its peak area of selected reaction monitoring (SRM) relative to that of the internal standard (N1-Ac-Spd-d6).

Metabolome Analysis
Capillary Electrophoresis Time-of-Flight MS and LC Time-of-Flight MS. Using capillary electrophoresis time-of-flight MS and LC time-of-flight MS with Advanced Scan Plus (Human Metabolome Technologies, Yamagata, Japan), comprehensive metabolome analysis was conducted based on methods described previously.16

LC-MS of Detailed Polyamine Metabolites. Spd, Spm, and their acetylated metabolites were separated by high-performance LC (HPLC; Gilson, Middleton, WI) using a Develosil ODS UG3 column (150 × 2.0mm, 3μm particle; Nomura Chemical Co, Aichi, Japan). LC conditions were modified from a method reported previously.22 Briefly, we used 0.1% HCOOH/0.05% heptfluorobutyric acid (HFBA)/H2O as mobile phase A and 0.1% HCOOH/0.05% HFBA/80% acetonitrile as mobile phase B, with a flow rate of 150μl/min. The HPLC system was connected to a TSQ Quantum Ultra AM mass spectrometer (Thermo Fisher Scientific, Waltham, MA). Target compounds were analyzed in the SRM positive-ionization mode.

Diffusion Tensor Imaging
Diffusion tensor images were acquired from 20 PD patients successively examined (3.06 ± 0.822 months after blood sampling) within Cohort B on a 3.0T system (Achieva; Philips Healthcare, Best, the Netherlands) with an 8-channel phased-array head coil for sensitivity encoding parallel imaging. Whole brain images were obtained using spin-echo echo planar imaging (EPI) with (1) 32 uniformly distributed direction diffusion-encoding (b = 1,000 s/mm2 for each direction), and (2) no diffusion weighting (b = 0 s/mm2) in an anterior–posterior phase-encoding direction. Standard- and reverse-phase encoded blipped images with no diffusion weighting were also acquired to correct for magnetic susceptibility-induced distortions related to EPI acquisitions.23 The scanning parameters were repetition time = 9,810 milliseconds, echo time = 100 milliseconds, field of view = 256 × 256mm, matrix size = 128 × 128, slice thickness = 2mm, number of slices = 75, orientation = axial, and scanning time = 6.50 minutes. Diffusion tensor imaging (DTI) datasets were checked visually in all 3 orthogonal views. No dataset had severe artifacts related to gross geometric distortions, bulk motion, or signal dropout. Data were corrected for susceptibility-induced geometric distortions, eddy current distortions, and intervolume subject motion using the EDDY and TOPUP toolboxes.23 Tensors were computed at each voxel by fitting a tensor model to diffusion-weighted images with b = 0 and 1,000 s/mm2. Once the tensor was estimated, fractional anisotropy (FA), mean diffusivity (MD), axial diffusivity (AD), and radial diffusivity (RD) were estimated based on standard formulae.24

DTI Image Processing Using Tract-Based Spatial Statistics Analysis
Whole-brain voxelwise analysis of all diffusion tensor–derived measurements was performed using tract-based spatial statistics (TBSS) implemented with FMRIB Software Library 5.0.9 (FSL, Oxford Centre for Functional MRI of the Brain, Oxford, UK; www.fmrib.ox.ac.uk/fsl).25 Briefly, FA images of all subjects were registered into FMRIB58_FA standard-space images with the nonlinear registration tool FNIRT, followed by visual inspection to ensure registration quality. Then, mean FA images were created by averaging registered FA images. Furthermore, mean FA images were thinned to generate a mean FA skeleton representing the centers of all tracts of groups. Aligned FA images from each subject were projected onto the mean FA skeleton using a lower threshold of FA = 0.2 to exclude peripheral tracts and gray matter. By applying the original nonlinear registration warping field of each subject’s FA to the standard space, MD, AD, and RD were also projected onto the mean FA skeleton. Furthermore, MD, AD, and RD data were also used to calculate voxelwise statistics. Anatomical locations of regions with significant correlations in the white-matter skeleton were identified by the Johns Hopkins University DTI white matter atlas within FSL.26 In addition, average diffusion metrics of clusters that showed significant correlation by TBSS analysis were measured.

Genomic DNA Analysis
DNA was extracted from peripheral blood using a Qiagen Kit (Qiagen, Venlo, the Netherlands). To rule out the possibility that Spd and/or Spm metabolism is changed by SAT2 or histone deacetylase 10 (HDAC10),27,28 genetic screening of SAT1, SAT2, and HDAC10 were performed. Eighteen single-nucleotide variants (SNVs) present in >1% of the Japanese population were
chosen based on the Integrative Japanese Genome Variation Database (ijGVGD). Of these, 16 SNVs (rs13894, rs858521, rs858520, and rs139435483 of the SAT2 gene; rs75596977, rs7771271, rs76578729, rs7096954, rs4838866, rs35820251, rs41283469, rs738334, rs1555048, rs150016700, rs77003572, and rs76662439 of the HDAC10 gene) were examined using the QuantStudio 7 Flex Real-Time PCR System (Thermo Fisher Scientific) with TaqMan genotyping assays (TaqMan Fisher Scientific). The final volume was 10 μl consisting of 5 μl TaqMan GTXpress Master Mix, 0.5 μl 2 mM L-glutamine, at 37 μl TaqMan genotyping assay, 2.5 μl nuclease-free water, and 2 μl genomic DNA in each well of a 96-well plate. Real-time polymerase chain reaction (PCR) steps included an initial enzyme activation step at 95°C for 20 minutes, followed by 40 cycles of a denaturation step at 95°C for 1 second and an annealing/extension step at 60°C for 3 seconds. The other SNVs (rs11553697 and rs2294404 of the HDAC10 gene) and all cording exons and exon-intron boundaries of SAT1, spermine synthase (SMS), and spermidine oxidase (SMOX) genes were sequenced using the Sanger method with BigDye Terminators v3.1 Cycle Sequencing Kit and 3130 Genetic Analyzer (Life Technologies, Foster City, CA) PCR, and sequencing primers were designed by Primer 3. Frequencies of analyzed SNVs were referred to gnomAD (https://gnomad.broadinstitute.org/) and compared with iJGVD. Of these, 16 SNVs were correlated with blood metabolite levels and LED or UPDRS-III in PD using JMP13. Conditional logistic regression analyses were used to calculate odds ratios and 95% confidence intervals for allelic and genotypic correlations with PD risk. Allele frequencies were analyzed with Pearson chi-squared test. Probability of p < 0.05 was considered statistically significant.

The Randomize tool (https://fsl.fmrib.ox.ac.uk/fsl/fslwiki/Randomise) was used to examine relationships between diffusion metrics and blood metabolites (DiAcSpd, N8-AcSpd, DiAcSpm, and Spm/Spd ratio) by multiple linear regression analysis. A general linear model was created for analysis of covariance (ANCOVA), with age and LED as nuisance covariates to adjust for their potential confounding influence on DTI measurements. To avoid selection of an arbitrary cluster-forming threshold, the threshold free cluster enhancement option was used in Randomize. In total, 5,000 permutations were generated to provide an empirical null distribution of maximal cluster size. For free water elimination, p < 0.05 was considered significant for TBSS. In addition, mean diffusion tensor–derived measurements of significant clusters were correlated with blood metabolites using Spearman rank correlation test, with significance defined at p < 0.05.

Results

Two Acetylation Forms, N8-AcSpd and N-Acetylputrescine, Correlated with PD Severity in Cohort A

Comprehensive metabolome analysis using Cohort A (named “2nd cohort” in our previous report) showed elevated N8-AcSpd (ratio of PD to control = 1.44, p = 0.0036) and N-acetylputrescine (N-AcPut; ratio of PD to control = 1.20, p = 0.126) in PD relative to healthy controls. Levels of the other polyamine metabolites were below the detection limit. Both N8-AcSpd and N-AcPut positively correlated with H&Y (N8-AcSpd, p < 0.0001; N-AcPut, p = 0.0002),
UPDRS-III ($N^8$-AcSpd, $p < 0.0001$; $N$-AcPut, $p = 0.0002$), age at sampling (hereafter simply referred to as “age”); $N^8$-AcSpd, $p = 0.0007$; $N$-AcPut, $p < 0.0001$), and LED ($N^8$-AcSpd, $p = 0.0018$; $N$-AcPut, $p < 0.0001$). Age and LED strongly correlated with H&Y (age, $p = 0.0053$; LED, $p < 0.0001$). Thus, to exclude interactions of LED or age with both acetylated polyamines, we performed multiple regression analyses and found that $N^8$-AcSpd significantly correlated with H&Y (LED, $p < 0.0001$) and mildly with UPDRS-III (LED, $p = 0.0018$). Age and LED strongly correlated with H&Y (age, $p = 0.0007$; LED, $p < 0.0001$). Thus, to exclude interactions of LED or age with both acetylated polyamines, we performed multiple regression analyses and found that $N^8$-AcSpd significantly correlated with H&Y (LED, $p < 0.0001$) and mildly with UPDRS-III (LED, $p = 0.0018$). Meanwhile, $N$-AcPut mildly correlated with H&Y (LED, $p = 0.0667$) and UPDRS-III (LED, $p = 0.0932$). However, $N^8$-AcSpd and $N$-AcPut significantly correlated with H&Y (age, $p < 0.0001$ and $p = 0.0036$, respectively) and UPDRS-III (age, $p = 0.0004$ and $p = 0.0042$, respectively).

**DiAcSpd Correlates with Severity of PD**

Further exploration of polyamine metabolic changes in PD incorporated a larger validation cohort (Cohort B), including tauopathy disease controls (PSP and AD; Table 1, Cohort B). In Cohort B, we performed serum analysis, as it enabled us to perform more sensitive analysis of Spm and Spd with similar sensitivity to their acetylated forms compared with plasma analysis (data not shown). As expected, most acetylated polyamine forms were significantly elevated in PD (Table 2). Importantly, Spd was significantly higher in PD than controls, whereas Spm was significantly lower in PD. A similar tendency in levels of each polyamine metabolite was also detected in 4 de novo PD patients.

As shown in Figure 2A–G, DiAcSpd, $N^8$-AcSpd, and DiAcSpm positively correlated with H&Y, UPDRS-III

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**TABLE 1. Demographic Characteristics of Participants in Both Cohorts**

| Characteristic          | Cohort A | PD | Cohort B | PD | $p$ (control vs PD)* | De Novo PD | PSP | AD |
|-------------------------|----------|----|----------|----|----------------------|------------|-----|----|
| n                       | 45       | 145| 49       | 186| —                    | 4          | 19  | 23 |
| Sex, M:F                | 23:22    | 70:75| 24:25    | 80:106| 0.455b               | 1.3        | 13:6| 7:16 |
| Age, mean yr (SD)       | 63.8 (15.3) | 67.5 (10.2) | 63.0 (14.7) | 67.6 (9.56) | 0.105      | 68.5 (5.45) | 71.0 (6.86) | 75.2 (8.96) |
| Duration, mean yr (SD)c | —        | 7.04 (5.61) | —        | 7.27 (5.36) | —          | 1.75 (0.957) | 4.89 (2.26) | 3.61 (2.21) |
| H&Y stage, mean (SD)    | —        | 2.09 (0.897) | —        | 2.05 (0.907) | —          | 2.00 (0.00) | 3.76 (1.15) | —   |
| H&Y stage (cases, n)    | —        | I (41), II (60), III (35), IV (8), V (1) | —        | 0 (1), I (54), II (81), III (35), IV (15), V (6) | —          | II (4) | I (1), II (2), III (3), IV (9), V (4) | —   |
| UPDRS-III, mean (range) | —        | 14.8 (1–57) | —        | 13.5 (0–54) | —          | 10.0 (8–12) | —   | —   |
| MMSE, mean (SD)         | 28.9 (2.09) | 27.8 (3.14) | 28.8 (2.24) | 27.8 (3.10) | 0.021      | 27.0 (0.00) | 20.3 (9.54) | 18.2 (6.45) |
| BMI, mean kg/m² (SD)    | 23.2 (3.59) | 22.4 (3.29) | 23.2 (3.69) | 22.6 (3.37) | 0.184      | 21.1 (0.518) | —   | —   |
| L-dopa, mg              | —        | 386 (214) | —        | 399 (239) | —          | 0          | 371 (302) | —   |
| L-dopa, equivalent dose | —        | 618 (342) | —        | 606 (357) | —          | 0          | 494 (343) | —   |

Cohort A corresponds to 2nd cohort in our previous report.16

*Probability values obtained by Wilcoxon test between controls and PD within Cohort B.

*bProbability value obtained by chi-squared test between controls and PD within Cohort B.

*cDuration was defined as time since onset of initial motor symptoms.

AD = Alzheimer disease; BMI = body mass index; F = female; H&Y = Hoehn and Yahr stage; M = male; MMSE = Mini-Mental State Examination; PD = Parkinson disease; PSP = progressive supranuclear palsy; SD = standard deviation; UPDRS-III = Unified Parkinson’s Disease Rating Scale motor section.
DiAcSpd, $p = 0.0023$; N8-AcSpd, $p = 0.0046$; DiAcSpm, $p = 0.0103$), and LED (DiAcSpd, $p = 0.0043$; N1-AcSpd, $p = 0.0029$; DiAcSpm, $p = 0.00223$). ROC curve analysis with these 7 metabolites showed a high diagnostic value (see Fig 2H) similar to DiAcSpd (AUC = 0.946, cutoff value = 0.821). Furthermore, these polyamine metabolites had robust diagnostic power for differentiating PD from PSP (AUC = 0.931, cutoff value = 0.805) and AD (AUC = 0.938, cutoff value = 0.801).

As in Cohort A, age, disease duration and LED positively correlated with H&Y stage (age, $p < 0.0001$; disease duration, $p < 0.0001$; LED, $p < 0.0001$); therefore, logistic regression and multiple regression analyses were performed to exclude their influence for accurate assessment of the relationship between acetylated polyamines and disease severity. Positive correlations of disease severity with DiAcSpd were detected under normalization of both LED (Table 3) and age (Table 4). As in Cohort A, age, disease duration and LED positively correlated with H&Y stage (age, $p < 0.0001$; disease duration, $p < 0.0001$; LED, $p < 0.0001$); therefore, logistic regression and multiple regression analyses were performed to exclude their influence for accurate assessment of the relationship between acetylated polyamines and disease severity. Positive correlations of disease severity with DiAcSpd were detected under normalization of both LED (Table 3) and age (Table 4). As in Cohort A, age, disease duration and LED positively correlated with H&Y stage (age, $p < 0.0001$; disease duration, $p < 0.0001$; LED, $p < 0.0001$); therefore, logistic regression and multiple regression analyses were performed to exclude their influence for accurate assessment of the relationship between acetylated polyamines and disease severity. Positive correlations of disease severity with DiAcSpd were detected under normalization of both LED (Table 3) and age (Table 4). As in Cohort A, age, disease duration and LED positively correlated with H&Y stage (age, $p < 0.0001$; disease duration, $p < 0.0001$; LED, $p < 0.0001$); therefore, logistic regression and multiple regression analyses were performed to exclude their influence for accurate assessment of the relationship between acetylated polyamines and disease severity. Positive correlations of disease severity with DiAcSpd were detected under normalization of both LED (Table 3) and age (Table 4). As in Cohort A, age, disease duration and LED positively correlated with H&Y stage (age, $p < 0.0001$; disease duration, $p < 0.0001$; LED, $p < 0.0001$); therefore, logistic regression and multiple regression analyses were performed to exclude their influence for accurate assessment of the relationship between acetylated polyamines and disease severity. Positive correlations of disease severity with DiAcSpd were detected under normalization of both LED (Table 3) and age (Table 4).

To confirm that these polyamine metabolites reflect disease severity of PD, we performed DTI on 20 patients with PD who were successively examined within Cohort B. As shown in Table 5, no significant differences were observed for any clinical characteristics and polyamine metabolite levels between PD patients and PD patients with DTI. Using DTI, we investigated association between levels of the 3 acetylated polyamines and axonal changes detected in PD.\textsuperscript{34} Accordingly, TBSS analysis detected a significant positive correlation between DiAcSpd and FA in an extensive white matter area in the brain of PD patients after normalization of LED and age (see Fig 2I, J; $p < 0.05$, familywise error-corrected; Peak Montreal Neurological Institute method x, y, z: 77, 90, 110; $t_{\text{max}} = 7.06$; $r_{\text{max}} = 0.70$; voxels = 45,836).\textsuperscript{35} Spearman rank correlation test also revealed that mean FA values for significant clusters in PD patients correlated positively with DiAcSpd ($r = 0.63$, $p = 0.003$).

### Analysis of Polyamine Metabolism-Associated Genes

Next, we genotyped 19 SNVs to investigate whether variants of 3 genes (\textit{SAT1}, \textit{SAT2}, and \textit{HDAC10}) that encode enzymes associated with synthesis or acetylation/deacetylation of Spd and Spm affect polyamine metabolism in patients with PD. In addition, Sanger sequencing for mutation detection in the \textit{SAT1}, \textit{SMS}, and \textit{SMOX} genes was performed. We did not detect any variants with statistically different frequencies between patients with PD and controls (Supplementary Table).\textsuperscript{29}

### Spm/Spd Ratio Is Consistently Suppressed in PD Independent of Age

The Spm/Spd ratio, indicating Spm conversion from Spd, significantly decreased in the PD group compared with the healthy control group, as a whole (ratio to control = 0.459) and in each of the H&Y stages as assessed by Steel test (ratio to control: I = 0.451, II = 0.470, III = 0.466, IV = 0.447; Fig 3). The Spm/Spd ratio enabled us to distinguish patients with PD from healthy controls. However, more distinct

### Table 2. Polyamine Metabolites in Cohort B Patients

| Compound Name | Comparative Analysis | PD/Control | De Novo PD/Control | PSP/Control | AD/Control |
|---------------|----------------------|------------|--------------------|------------|------------|
|               | Ratio p | Ratio p | Ratio p | Ratio p | Ratio p |
| DiAcSpd       | 2.77 <0.0001 | 1.95 0.134 | 1.93 <0.0001 | 1.51 0.0811 |
| N1-AcSpd      | 1.46 <0.0001 | 0.850 0.661 | 1.74 0.0002 | 1.11 0.237 |
| N8-AcSpd      | 1.55 <0.0001 | 1.09 0.968 | 1.89 <0.0001 | 1.53 0.0006 |
| DiAcSpm       | 1.59 <0.0001 | 0.892 1.00 | 1.65 0.0001 | 2.11 0.0763 |
| Spd           | 1.80 <0.0001 | 1.57 0.0764 | 1.05 1.00 | 1.02 0.886 |
| N1-AcSpm      | 0.945 0.302 | 0.770 0.404 | 1.43 0.0012 | 1.05 0.999 |
| Spm           | 0.762 0.0468 | 0.776 0.951 | 1.21 0.223 | 0.563 0.0013 |
| Spm/Spd ratio | 0.459 <0.0001 | 0.515 0.106 | 1.43 0.154 | 0.544 <0.0001 |

Probability values were obtained by Steel test between healthy controls and each disease. AD = Alzheimer disease; DiAcSpd = N1,N8-diacetylspermidine; DiAcSpm = N1,N12-diacetylspermine; N1-AcSpd = N1-acetylspermidine; N1-AcSpm = N1-acetylspermine; N8-AcSpd = N8-acetylspermidine; PD = Parkinson disease; PSP = progressive supranuclear palsy; Spd = spermidine; Spm = spermine.

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\(\text{DiAcSpd}, p = 0.0023\); \(\text{N8-AcSpd}, p = 0.0046\); \(\text{DiAcSpm}, p = 0.0103\), and LED (\(\text{DiAcSpd}, p = 0.0043\); \(\text{N8-AcSpd}, p = 0.0029\); \(\text{DiAcSpm}, p = 0.00223\)). ROC curve analysis with these 7 metabolites showed a high diagnostic value (see Fig 2H) similar to DiAcSpd (AUC = 0.946, cutoff value = 0.821). Furthermore, these polyamine metabolites had robust diagnostic power for differentiating PD from PSP (AUC = 0.931, cutoff value = 0.805) and AD (AUC = 0.938, cutoff value = 0.801).

As in Cohort A, age, disease duration and LED positively correlated with H&Y stage (age, $p < 0.0001$; disease duration, $p < 0.0001$; LED, $p < 0.0001$); therefore, logistic regression and multiple regression analyses were performed to exclude their influence for accurate assessment of the relationship between acetylated polyamines and disease severity. Positive correlations of disease severity with DiAcSpd were detected under normalization of both LED (Table 3) and age (Table 4). Likewise, ANCOVA showed that the levels of 3 acetylated polyamines were significantly elevated in association with age in PD relative to controls (DiAcSpd, $p = 0.0056$; N8-AcSpd, $p = 0.0018$; DiAcSpm, $p = 0.0468$; DiAcSpm, $p = 0.0018$), which was partially consistent with a previous report.\textsuperscript{12}

To confirm that these polyamine metabolites reflect disease severity of PD, we performed DTI on 20 patients with PD who were successively examined within Cohort B. As shown in Table 5, no significant differences were observed for any clinical characteristics and polyamine metabolite levels between PD patients and PD patients with DTI. Using DTI, we investigated association between levels of the 3 acetylated polyamines and axonal changes detected in PD.\textsuperscript{34} Accordingly, TBSS analysis detected a significant positive correlation between DiAcSpd and FA in an extensive white matter area in the brain of PD patients after normalization of LED and age (see Fig 2I, J; $p < 0.05$, familywise error-corrected; Peak Montreal Neurological Institute method x, y, z: 77, 90, 110; $t_{\text{max}} = 7.06$; $r_{\text{max}} = 0.70$; voxels = 45,836).\textsuperscript{35} Spearman rank correlation test also revealed that mean FA values for significant clusters in PD patients correlated positively with DiAcSpd ($r = 0.63$, $p = 0.003$).
FIGURE 2: Acetylated spermidine (Spd)/spermine (Spm) positively correlates with Parkinson disease severity. (A–F) Levels of N1,N8-diacetylspermidine (DiAcSpd), N8-acetylspermidine (N8-AcSpd), and N1,N12-diacetylspermine (DiAcSpm) positively correlated with Parkinson disease severity assessed by Hoehn and Yahr stages (H&Y), whereas other polyamine metabolites did not. Values indicate the amount of each metabolite (pmol) in 200μl serum. (G) Levels of Spm were significantly decreased in Parkinson disease, at H&Y stages I and IV. Values indicate the amount of each metabolite (pmol) in 200μl serum. (H) Receiver operating characteristic curves for all serum polyamine metabolites and corresponding area under the curve (AUC) statistic for the true positive rate of Parkinson disease diagnosis in Cohort B. (I) Scatterplots showing positive relationships between DiAcSpd and fractional anisotropy (FA) in Parkinson disease. (J) Significant positive correlation between N1,N8-diacetylspermidine and FA, adjusted for age and L-dopa equivalent dose (p < 0.05, familywise error corrected) in Parkinson disease patients. Significant clusters were overlaid onto a Montreal Neurological Institute (MNI) ICBM152 standard brain T1-weighted image. Slices in MNI coordinates x, y, z are shown in millimeters. Colored bar represents p value. **p < 0.01, ***p < 0.001 (Steel test or analysis of variance). N1-AcSpd = N1-acetylspermidine; N1-AcSpm: N1-acetylspermine; NS = not significant.
classification was provided using both Spm/Spd ratio and DiAcSpd. Interestingly, PD patients showed significantly lower Spm/Spd ratios at any age ($p = 0.0007$), whereas healthy controls showed higher ratios especially from 40 to 60 years old. This negative correlation with age is compatible with reports showing age-dependent decreases in Spm levels. Altogether, these data suggest that decreased Spm effects with aging might be associated with onset risk of PD.

### TABLE 3. Correlation Analysis with Logistic Regression or Multiple Regression Model of Each Polyamine Metabolite, Disease Severity (H&Y or UPDRS-III), and LED

| Compound Name | H&Y (normalized to LED) | UPDRS-III (normalized to LED) |
|---------------|-------------------------|-------------------------------|
|               | $F^a$ | $p^a$ | $F^b$ | $p^b$ |
| DiAcSpd       | 3.45  | 0.0096| 4.28  | 0.04  |
| $N1$-AcSpd    | 0.77  | 0.546 | 0.169 | 0.681 |
| $N8$-AcSpd    | 3.08  | 0.0176| 3.08  | 0.0808|
| DiAcSpm       | 3.73  | 0.0061| 3.2   | 0.0755|
| Spd           | 1.31  | 0.27  | 1.46  | 0.229 |
| $N1$-AcSpm    | 1.22  | 0.303 | 0.576 | 0.449 |
| Spm           | 1.62  | 0.172 | 0.0022| 0.962 |
| Spm/Spd ratio | 0.156 | 0.96  | 1.15  | 0.285 |

*Obtained by logistic regression analysis.
*Obtained by multiple regression analysis.

DiAcSpd = $N1,N8$-diacetylspermidine; DiAcSpm = $N1,N12$-diacetylspermine; H&Y = Hoehn and Yahr stage; LED = L-dopa equivalent dose; $N1$-AcSpd = $N1$-acetylspermidine; $N1$-AcSpm = $N1$-acetylspermine; $N8$-AcSpd = $N8$-acetylspermidine; Spd = spermidine; Spm = spermine; UPDRS-III = Unified Parkinson’s Disease Rating Scale motor section.

### TABLE 4. Correlation Analysis with Logistic Regression or Multiple Regression Model of Each Polyamine Metabolite, Disease Severity (H&Y or UPDRS-III), and Age at Sampling

| Compound Name | H&Y (normalized by age at sampling) | UPDRS-III (normalized by age at sampling) |
|---------------|-----------------------------------|------------------------------------------|
|               | $F^a$ | $p^a$ | $F^b$ | $p^b$ |
| DiAcSpd       | 3.18  | 0.015 | 5.85  | 0.0165|
| $N1$-AcSpd    | 1.32  | 0.263 | 0.146 | 0.228 |
| $N8$-AcSpd    | 3.22  | 0.014 | 4.58  | 0.0337|
| DiAcSpm       | 2.94  | 0.0301| 3.26  | 0.0725|
| Spd           | 1.45  | 0.219 | 1.37  | 0.244 |
| $N1$-AcSpm    | 1.41  | 0.233 | 0.258 | 0.612 |
| Spm           | 1.79  | 0.132 | 0.730 | 0.394 |
| Spm/Spd ratio | 0.185 | 0.946 | 0.0407| 0.840 |

*Obtained by logistic regression analysis.
*Obtained by multiple regression analysis.

DiAcSpd = $N1,N8$-diacetylspermidine; DiAcSpm = $N1,N12$-diacetylspermine; H&Y = Hoehn and Yahr stage; $N1$-AcSpd = $N1$-acetylspermidine; $N1$-AcSpm = $N1$-acetylspermine; $N8$-AcSpd = $N8$-acetylspermidine; Spd = spermidine; Spm = spermine; UPDRS-III = Unified Parkinson’s Disease Rating Scale motor section.
Spm at 5 to 10 μM, and 10 to 50 μM Spd or N1-AcSpm Induce Autophagy in SH-SY5Y Cells

Because both Spm and the Spm/Spd ratio were significantly decreased in PD, we speculated that autophagic activity, an aging-modulating system, may be influenced by polyamine metabolic changes in PD. Thus, we investigated Spm effects on various cell lines using the LC3-II/β-actin ratio. Spm upregulated the LC3-II/β-actin ratio drastically in human neuroblastoma SH-SY5Y cells, and moderately in human osteosarcoma U2OS cells (Fig 4A). Next, the effects of 3 polyamines and 5 acetylated forms on autophagy in SH-SY5Y cells were examined. Of these 8 molecules, Spm, N1-AcSpd, and Spd significantly upregulated LC3-II levels (Fig 4B). Quantitative analysis of LC3-II levels in 3 independent samples showed that Spm upregulated at 5 μM, whereas N1-AcSpm and Spd modulated at 25 μM (Fig 4C).

LC3-II upregulation is induced by autophagosome formation or impairment of autophagosome maturation. In the presence of bafilomycin A1, an inhibitor of lysosomal acidification and of autophagosome–lysosome fusion, Spm and Spd increased LC3-II levels compared with controls (see Fig 4D). Similar results were obtained when cells were cotreated with the lysosomal protease inhibitors pepstatin A and E64D (see Fig 4E), indicating that Spm and Spd induce autophagy, and that Spm is a more potent autophagy inducer. These data imply that decreased levels of Spm/Spd may affect autophagic activity in PD patients.

Discussion

In the current study, we identified DiAcSpd as a diagnostic, severity-associated, medication-independent biomarker of PD, although gender distribution differences between controls and PD may affect polyamine metabolism. Notably, even using a multivariate model, DiAcSpd remained significantly correlated with disease severity. In addition, decreases in Spm and the Spm/Spd ratio were detected in PD and AD, but not control or PSP groups. In particular, the Spm/Spd ratio was consistently decreased across ages in PD, unlike the ratio in controls, which showed gradual decrements, suggesting a PD-associated aging tendency. Levels of Spd were only mildly increased in de novo PD and under medication-normalized conditions. Finally, we confirmed that among the polyamine metabolites tested, Spm enhanced autophagy most intensely in SH-SY5Y cells, which were used...
as a neuronal model for PD and AD, implying a decrement in autophagic activity in PD and AD.

We and others have reported serum/plasma metabolite biomarkers for early diagnosis of PD. However, surrogate biomarkers reflecting age-related pathogeneses have not been established.\(^1^6,36^\text{-}38\) DiAcSpd levels clearly correlated with both H&Y and UPDRS-III under age- or medication-normalized conditions, and significantly correlated with FA (an index of white matter integrity alterations), in diffuse white matter, similar to our previous report.\(^34,39\) Likewise, higher diagnostic power of DiAcSpd levels was confirmed. Taken together, serum DiAcSpd measurements could be a potential diagnostic biomarker correlating with disease severity.

Several lines of evidences have shown changes in CSF polyamine levels in patients with various diseases including brain tumors, inflammation, and neurodegeneration.\(^7\) In

FIGURE 3: Conversion of spermidine (Spd) to spermine (Spm) is suppressed in Parkinson disease (PD), without association with disease severity. (A) Sm/Spd ratio was significantly decreased in PD patients. ***p < 0.001 (Wilcoxon test). (B) Multiple comparisons of the Sm/Spd ratios showed significant decreases in each Hoehn and Yahr stage (H&Y) relative to the controls. ***p < 0.001 (Steel test). (C) Receiver operating characteristic (ROC) curve analysis of Sm/Spd ratio. (D) ROC curve analysis of Sm/Spd ratio and N1,N8-diacetylspermidine. (E) Sm/Spd ratios of all controls and PD patients shown on the same graph. Interaction was assessed by analysis of covariance (ANCOVA) between control and PD groups. AUC = area under the curve.
PD compared with controls, Spd decreases in CSF, whereas putrescine increases in CSF and decreases in red blood cells, and no significant changes in Spd and Spm in the basal ganglia have been reported. Concentrations of Spd and Spm in the basal ganglia gradually decrease in normal aging. Experimentally, Spm clearance from the brain parenchyma mainly occurs via the blood–CSF (BCSFB) rather than the blood–brain barrier. Taken together, decreased serum Spm detected in PD in the current study may indicate that brain Spm decreases via the BCSFB.

Compared with PD, YKL-40 levels in CSF are increased in PSP, which is often difficult to diagnose accurately, particularly in the early stages. No serum/plasma differential biomarkers have been established. In the current study, serum polyamine metabolite profile appears to be a potential diagnostic tool for differentiating PD from PSP and AD. Moreover, aging is the most important causative risk factor for PD, PSP, and AD. Mean age at onset of neurological symptoms of PSP is 66.4 ± 12 years, similar to PD, but disease prevalence of PSP does not correlate with aging. This is consistent with our finding that
Spm and the Spm/Spd ratio are suppressed in PD and AD, but that the ratio in PSP is not significantly different from controls.

Autophagy deficiency causes neurodegeneration along with aggregated protein accumulation in the cytoplasm.\textsuperscript{45,46} Notably, in PD, α-synuclein included in Lewy bodies is an autophagy substrate and targeted by disease-modifying therapy with autophagy-inducing chemicals.\textsuperscript{47} In addition, α-synuclein itself inhibits autophagic flux in cellular models.\textsuperscript{48} Growing evidence implies that Spd is involved in antiaging through induction of autophagy.\textsuperscript{6} Furthermore, oral Spd administration protects flies against neurodegeneration-like age-induced memory impairment and α-synuclein–induced motor dysfunction through autophagy enhancement, indicating that Spd-induced autophagy is indispensable for lifespan-extension/nerve protection.\textsuperscript{6} Antiaging effects of Spm have also been reported in mice.\textsuperscript{49} Because our data showed that Spm was 3 to 4 times more effective in autophagy induction than Spd at 5 to 20μM, which is within physiological blood concentrations, and considering decreased levels of Spm and Spm/Spd ratios in PD and AD, conversion from Spd to Spm may be crucial for the maintenance of homeostasis in neuronal cell lines.

This study has some limitations. First, it was conducted at a single university hospital. Second, not all enrolled patients underwent DTI scans. Finally, medications or cancers may affect polyamine metabolism. Although antiparkinsonian medications may influence polyamine metabolism, which concomitantly occurs in various tissues, the hyperacetylation status was confirmed under normalized condition of antiparkinsonian medications in our cohorts. Polyamine metabolism is upregulated in patients with urogenital or colorectal cancers; thus, we excluded patients with any known cancer in both cohorts.\textsuperscript{50} Future studies should test all polyamine-associated metabolites, including ornithine and putrescine, in larger cohorts. In addition, longitudinal observation in the same cases should be included in future research.

Herein, DiAcSpd is shown to be a potential, peripheral, noninvasive, diagnostic biomarker of PD that correlates with disease severity. In addition, we demonstrate that dysfunction in conversion from Spd to Spm is an age-related risk for PD. Our study provides new insight into the association between aging risk for PD and autophagic activity regulated by Spm synthesis from Spd.

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Author Contributions

Study concept and design: S.Sa., Y.S., N.H. Data acquisition and analysis: all authors. Drafting text: S.Sa., Y.S., N.H.

Potential Conflicts of Interest

Nothing to report.

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