

ORIGINAL RESEARCH

Open Access



# Impact of a combination of quantitative indices representing uptake intensity, shape, and asymmetry in DAT SPECT using machine learning: comparison of different volume of interest settings

Yu Iwabuchi<sup>1</sup>, Tadaki Nakahara<sup>1\*</sup>, Masashi Kameyama<sup>1,2</sup>, Yoshitake Yamada<sup>1</sup>, Masahiro Hashimoto<sup>1</sup>, Yohji Matsusaka<sup>1</sup>, Takashi Osada<sup>3</sup>, Daisuke Ito<sup>3</sup>, Hajime Tabuchi<sup>4</sup> and Masahiro Jinzaki<sup>1</sup>

## Abstract

**Background:** We sought to assess the machine learning-based combined diagnostic accuracy of three types of quantitative indices obtained using dopamine transporter single-photon emission computed tomography (DAT SPECT)—specific binding ratio (SBR), putamen-to-caudate ratio (PCR)/fractal dimension (FD), and asymmetry index (AI)—for parkinsonian syndrome (PS). We also aimed to compare the effect of two different types of volume of interest (VOI) settings from commercially available software packages DaTQUANT (Q) and DaTView (V) on diagnostic accuracy.

**Methods:** Seventy-one patients with PS and 40 without PS (NPS) were enrolled. Using SPECT images obtained from these patients, three quantitative indices were calculated at two different VOI settings each. SBR-Q, PCR-Q, and AI-Q were derived using the VOI settings from DaTQUANT, whereas SBR-V, FD-V, and AI-V were derived using those from DaTView. We compared the diagnostic value of these six indices for PS. We incorporated a support vector machine (SVM) classifier for assessing the combined accuracy of the three indices (SVM-Q: combination of SBR-Q, PCR-Q, and AI-Q; SVM-V: combination of SBR-V, FD-V, and AI-V). A Mann-Whitney *U* test and receiver-operating characteristics (ROC) analysis were used for statistical analyses.

**Results:** ROC analyses demonstrated that the areas under the curve (AUC) for SBR-Q, PCR-Q, AI-Q, SBR-V, FD-V, and AI-V were 0.978, 0.837, 0.802, 0.906, 0.972, and 0.829, respectively. On comparing the corresponding quantitative indices between the two types of VOI settings, SBR-Q performed better than SBR-V ( $p = 0.006$ ), whereas FD-V performed better than PCR-Q ( $p = 0.0003$ ). No significant difference was observed between AI-Q and AI-V ( $p = 0.56$ ). The AUCs for SVM-Q and SVM-V were 0.988 and 0.994, respectively; the two different VOI settings displayed no significant differences in terms of diagnostic accuracy ( $p = 0.48$ ).

**Conclusion:** The combination of the three indices obtained using the SVM classifier improved the diagnostic performance for PS; this performance did not differ based on the VOI settings and software used.

**Keywords:** <sup>123</sup>I-Ioflupane, <sup>123</sup>I-FP-CIT, DAT SPECT, Parkinson's syndrome, Support vector machine, Machine learning

\* Correspondence: [nakahara@rad.med.keio.ac.jp](mailto:nakahara@rad.med.keio.ac.jp)

<sup>1</sup>Department of Radiology, Keio University School of Medicine, 35 Shinanomachi, Shinjyuku-ku, Tokyo 160-8582, Japan

Full list of author information is available at the end of the article

## Background

Two different types of volume of interest (VOI) settings have been proposed for routine clinical quantification of dopamine transporter single-photon emission computed tomography (DAT SPECT) results using VOI-based analyses. One of the two VOI settings uses a striatal VOI delineated to fit the striatum precisely. Some software packages, such as DaTQUANT (GE Healthcare, Little Chalfont, UK), apply this type of VOI setting [1]. This type of VOI setting needs registration to the normalized SPECT image because the shape and position of the striatum vary among individuals. This fixed-size VOI template can be divided to conform with the striatal structure parts, i.e., the caudate, the anterior putamen, and the posterior putamen, to enable the calculation of the putamen-to-caudate ratio (PCR) that represents the ratio between anterior and posterior striatal radioisotope (RI) uptake [1, 2]. Tossici-Bolt et al. proposed a large pentagonal prism-shaped VOI, which encompasses a wide area around the striatum (Southampton method), as an alternative setting [3]. Some software packages, such as DaTView (AZE, Tokyo, Japan), apply this type of VOI setting. This VOI can be set directly on the original image without normalization and has the merit of reducing the influence of a partial volume effect as it is large in size.

These two VOI settings can be used to evaluate three different types of quantitative indices that have complementary roles: the specific binding ratio (SBR), PCR/fractal dimension (FD), and asymmetry index (AI). The SBR index represents the amount of RI accumulation in the striatum, the PCR/FD index represents the distribution or the shape of the RI uptake, and the AI index represents the lateral difference between the right and left side RI uptake, respectively. Since the VOI from Tossici-Bolt et al. cannot be separated into anterior and posterior parts, the PCR index cannot be calculated using this VOI analysis. Therefore, another quantitative index that can replace the PCR index is required. Iwabuchi et al. proposed the FD index as a new quantitative index for assessing DAT SPECT results [4]. It can be calculated with the VOI from Tossici-Bolt et al. The FD index is used to evaluate the spatial heterogeneity or distribution of various types of medical images [5]. We consider that the PCR and FD indices have similar roles in assessing DAT SPECT as both indices represent the shape or the distribution of striatal RI accumulation.

In visual interpretations, a comprehensive evaluation of the three characteristic findings is important, i.e., the decline of striatal RI accumulation, the shape of RI accumulation, and the lateral difference in RI accumulation [6]. This is because the normal striatum shows symmetrical comma-shaped regions with high accumulation in both the caudate and putamen, while asymmetric dot-shaped patterns are observed in patients with Parkinsonian syndrome (PS). Therefore, we hypothesized that, even with quantitative assessments, a comprehensive evaluation with three

characteristic indices, namely the SBR, PCR/FD, and AI, would be similarly important. Although several previous studies have demonstrated the utility of these quantitative indices individually [7–11] or assessed the combined accuracy of SBR and FD [4], to our knowledge, no study has reported the combined usefulness of the indices using the three features (SBR, PCR/FD, and AI).

Further, the VOI-setting style is thought to affect the diagnostic accuracy in quantitative assessment [12]. Although Shimizu et al. have compared the usefulness of the two types of VOI settings in distinguishing dementia with Lewy bodies (DLB) from Alzheimer's disease (AD) [12]; they examined only the SBR index and did not examine other indices. Tossici-Bolt et al. and Yin et al. have demonstrated the influence of VOI settings on the SBR index itself [13, 14]. However, their studies did not assess the diagnostic accuracy for PS. To our knowledge, no study has assessed the effect of VOI settings on diagnostic accuracy for PS so far.

The purpose of this study was to assess the combined diagnostic accuracy of three types of DAT SPECT quantitative indices—SBR, PCR/FD, and AI—for PS using a support vector machine (SVM) classifier, and to compare the effect of two different types of VOI settings from commercially available software packages, i.e., DaTQUANT (Q) and DaTView (V).

## Materials and methods

### Patients

In this retrospective study, 183 consecutive patients who underwent DAT SPECT from February 2014 to August 2015 in our hospital were selected. Of these patients, 72 were excluded from the study because of the insufficiency of the image quality ( $n = 9$ ) or clinical diagnosis ( $n = 63$ ). The remaining 111 patients (median age, 68.5 years; range, 30–89 years; men/women, 62/49), including 71 patients with PS and 40 without PS (NPS), were included. Of the 71 patients with PS, 51 patients were diagnosed with PD based on the clinical diagnostic criteria of the UK Parkinson's disease society brain bank, and the other 20 patients were diagnosed with atypical parkinsonian syndrome including clinical DLB ( $n = 11$ ), progressive supranuclear palsy (PSP;  $n = 7$ ), and multiple system atrophy (MSA;  $n = 2$ ) on the basis of established diagnostic criteria [15–17].

The institutional review board of the hospital granted permission for this retrospective review of imaging and clinical data and waived the requirement for obtaining informed consent from the patients.

### SPECT acquisition and reconstruction

All SPECT images were acquired 3 h after the injection of  $^{123}\text{I}$ -Ioflupane (185 MBq) using the Discovery NM 630 or Discovery NM/CT 670 (GE Healthcare, Milwaukee, WI) with a FAN beam collimator. The imaging parameters were as follows: matrix size, 128 × 128; pixel size, 4.4 mm;

slice thickness, 4.4 mm; and energy window, 159 keV  $\pm$  10%. The projection data acquired for 30 min were reconstructed on a Xeleris workstation (GE Healthcare). The ordered-subset expectation maximization (OSEM) method (iterations, 3; subset, 10) and a Butterworth filter setting with a critical frequency of 0.5 and power of 10.0 were applied for DaTView software. The OSEM (iterations, 2; subset, 10) and a Butterworth filter setting with a critical frequency of 0.6 and power of 10.0 were applied for DaTQUANT. Neither attenuation correction nor scatter correction was used. These settings were determined based on the manufacturer's recommended settings.

#### Quantitative indices calculated from VOI-based analyses

We used two commercially available software packages for VOI-based analyses: DaTQUANT (GE Healthcare, Little Chalfont, UK) and DaTView (AZE, Tokyo, Japan) (Fig. 1). These are representative of two different widespread VOI-based approaches. DaTView uses the VOI from Tossici-Bolt et al., while DaTQUANT uses a normalized VOI template. The VOI settings for the reference background also differ. DaTView applies the whole brain, except a region around the basal ganglia, as a reference region, while DaTQUANT applies the occipital lobe as a reference region [1, 3].

The three types of indices derived with the VOI setting of DaTQUANT were named SBR-Q, PCR-Q, and AI-Q, while the indices derived with the VOI setting of DaTView were named SBR-V, FD-V, and AI-V. Five indices (SBR-Q, PCR-Q, AI-Q, SBR-V, and AI-V), except for FD-V, can be computed automatically with DaTQUANT and DaTView software. SBR-Q is defined as the mean counts of the striatal VOI (background-subtracted) divided by the mean counts of the occipital lobe VOI, which simply represents the count ratio [1]. On the other hand, SBR-V is defined as the count concentration of the striatal VOI divided by the count concentration of the whole brain, which represents

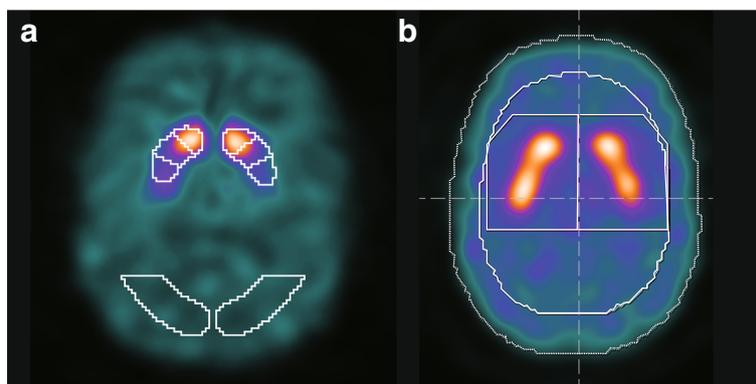
the concentration ratio [3]. Both the AI-Q and AI-V are defined as the difference between the SBR of both sides divided by the mean of the SBR of both sides. PCR-Q is defined as the mean counts of the putamen VOI divided by the mean counts of the caudate VOI [1]. The FD value can be calculated using the VOI setting from Tossici-Bolt et al. After setting 5-stage threshold values to delineate the striatum, the number of pixels obtained at each threshold was measured [4, 18–22]. The FD is calculated on the basis of the relationship between the threshold values and the number of pixels. For details, refer to a previous report [4].

#### Support vector machine classifier

We used SVM classification, one of the classifiers for machine learning, to assess the combined diagnostic accuracy of the three quantitative indices. SVM classifiers have been used for classification analyses owing to their high accuracy and ability to deal with high dimensional data. The SVM algorithm computes the class separation boundaries to maximize the distance between the boundaries and the example points belonging to different classes [23, 24]. In this study, linear SVM classifiers were used and the separation hyperplanes were defined using the following formula (1), where  $\mathbf{W}$  is the weight vector,  $X$  is the feature vector, and  $b$  is the bias.

$$\mathbf{W} \times X + b = 0 \quad (1)$$

We divided the dataset in such a way that 75% was used as a training dataset, and the remaining 25% was used as the testing data. This partitioning was stratified so that both datasets had the same class proportions. A tenfold cross validation was used for the training dataset to prevent overfitting [25]. The optimal  $\mathbf{W}$  and  $b$  that best separate the PS and NPS groups were computed using the training dataset. Furthermore, we calculated



**Fig. 1** Two types of VOI settings using DaTQUANT and DaTView. **a** DaTQUANT uses separated VOIs delineated to fit to the striatum. The reference VOIs are set to the bilateral occipital lobes. **b** DaTView uses large pentagonal prism-shaped VOIs encompassing a wide area around the striatum. The reference VOI is set to the whole brain. Abbreviation: VOI, volume of interest

**Table 1** Patient characteristics

	ALL	PS	NPS	<i>p</i> value
Number of cases	111	71	40	–
Age (year, mean ± SD)	69 ± 11.6	70 ± 10.2	66 ± 13.5	0.12*
Men/women ( <i>N</i> )	62/49	39/32	23/17	0.84†

No significant differences were observed with respect to age and the sex ratio between the PS and NPS groups

Abbreviations: PS parkinsonian syndrome, NPS non-parkinsonian syndrome, SD standard deviation

\**t* test

†Fisher's exact test

the diagnostic accuracy of the test dataset using the trained SVM classifier. We incorporated the three types of indices as features for SVM classification: SVM-Q (derived using the SBR-Q, PCR-Q, and AI-Q) and SVM-V (derived using the SBR-V, FD-V, and AI-V). Each index was incorporated after standardization using the following formula (2), where  $x$  is the original feature vector,  $\bar{x}$  is the mean of that feature vector, and  $\sigma$  is its standard deviation.

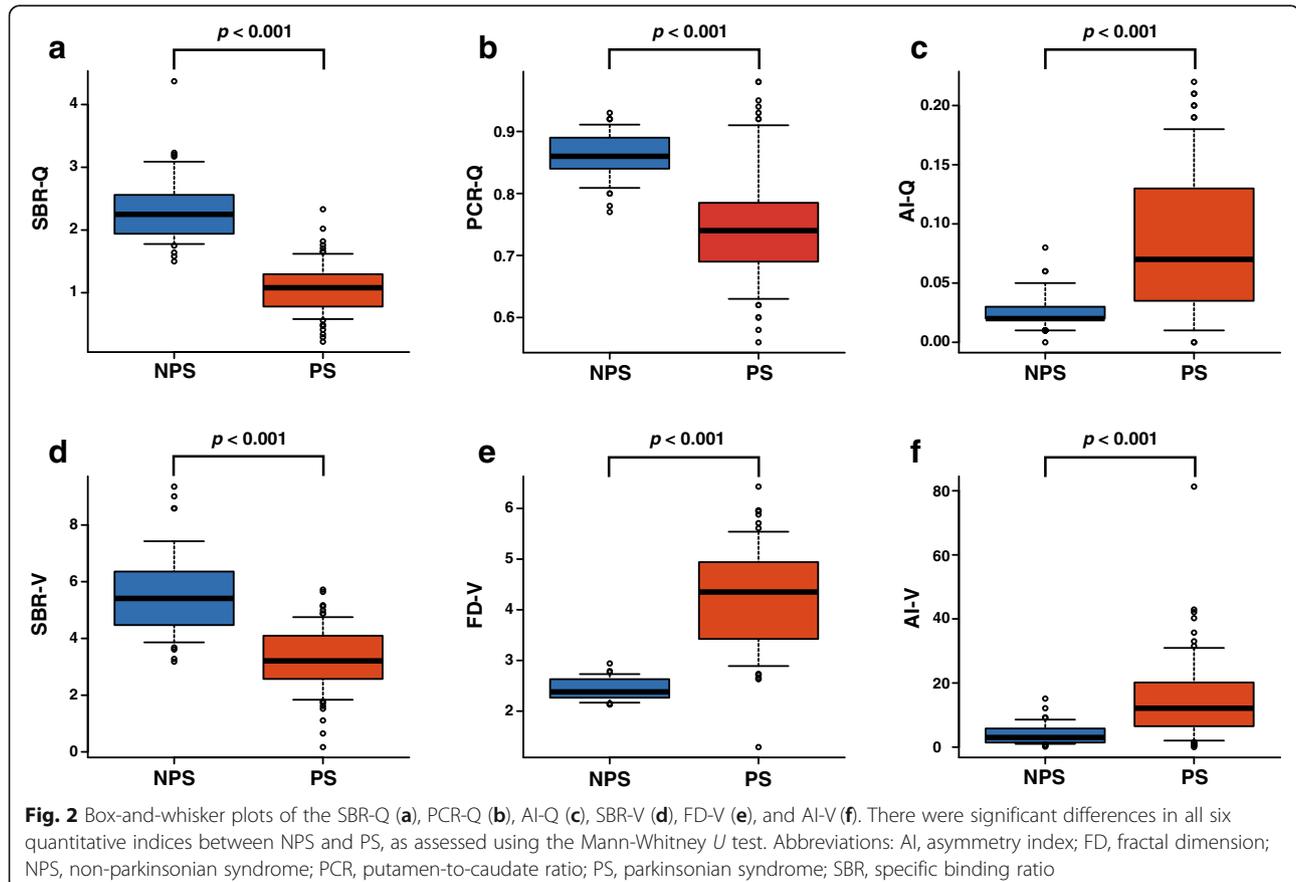
$$x' = \frac{x - \bar{x}}{\sigma} \quad (2)$$

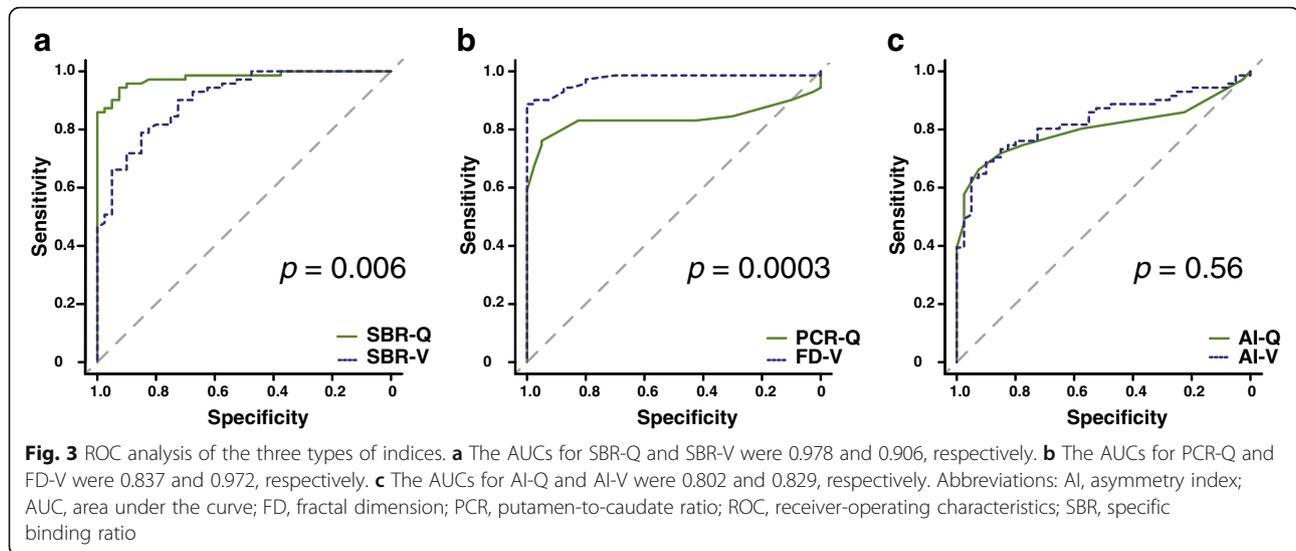
We compared the diagnostic accuracy between SVM-Q and SVM-V to determine the combined diagnostic accuracy of the two types of VOI settings.

### Statistical analysis

Sex and age distribution were compared using the Fisher's exact test or the *t* test. The Mann-Whitney *U* test was used to compare eight indices (SBR-Q, PCR-Q, AI-Q, SVM-Q, SBR-V, FD-V, AI-V, and SVM-V) between the PS and NPS groups. The receiver-operating characteristics (ROC) analysis of each index was performed to evaluate the AUC. We used the DeLong method to examine the difference between the two AUCs [26]. The sensitivity, specificity, positive predictive value (PPV), negative predictive value (NPV), and accuracy of each index were calculated using the optimal cutoff values determined based on the ROC curves. Differences with *p* values < 0.05 were considered statistically significant.

All statistical analyses were performed using the statistical package R (version 3.2.2; available as a free download from <http://www.r-project.org>) and SPSS software (version 14.0; SPSS Inc., Chicago, IL).





## Results

### Demographic features

The characteristics of the selected patients are summarized in Table 1. There were no significant group differences in sex and age.

### Comparison among individual quantitative indices

Figure 2 shows the box-and-whisker plots of the six quantitative indices used to differentiate between the PS and NPS groups. Significant differences were observed between the PS and NPS groups for every index ( $p < 0.001$ ). The mean SBR-Q (Fig. 2a), SBR-V (Fig. 2d), and PCR-Q (Fig. 2b) of patients with NPS were significantly higher than those of patients in the PS group ( $2.32 \pm 0.56$  vs  $1.07 \pm 0.42$ ,  $p < 0.001$ ;  $5.61 \pm 1.51$  vs  $3.26 \pm 1.15$ ,  $p < 0.001$ ;  $0.86 \pm 0.04$  vs  $0.75 \pm 0.10$ ,  $p < 0.001$ ; respectively), while the mean AI-Q (Fig. 2c), AI-V (Fig. 2f), and FD-V (Fig. 2e) were significantly lower for the NPS than for the PS group ( $0.03 \pm 0.02$  vs  $0.09 \pm 0.06$ ,  $p < 0.001$ ;  $3.94 \pm 3.37$  vs  $15.2 \pm 13.3$ ,  $p < 0.001$ ;  $2.44 \pm 0.21$  vs  $4.19 \pm 1.06$ ,  $p < 0.001$ ; respectively).

The ROC analyses demonstrated that the AUCs for SBR-Q, PCR-Q, AI-Q, SBR-V, FD-V, and AI-V were 0.978,

0.837, 0.802, 0.906, 0.972, and 0.829, respectively (Fig. 3). On comparing the corresponding quantitative indices between the two software packages, SBR-Q performed better than SBR-V ( $p = 0.006$ ), whereas FD-V performed better than PCR-Q ( $p = 0.0003$ ). No significant difference was observed between AI-Q and AI-V ( $p = 0.56$ ). Of the six indices, SBR-Q and FD-V showed better diagnostic performance than did other quantitative indices (SBR-Q, AUC = 0.978; FD-V, AUC = 0.972). Table 2 provides a summary of the sensitivity, specificity, PPV, NPV, and accuracy of the indices. The cutoff values for SBR-Q, PCR-Q, AI-Q, SBR-V, FD-V, and AI-V were 1.71, 0.81, 0.05, 4.16, 2.89, and 7.50, respectively.

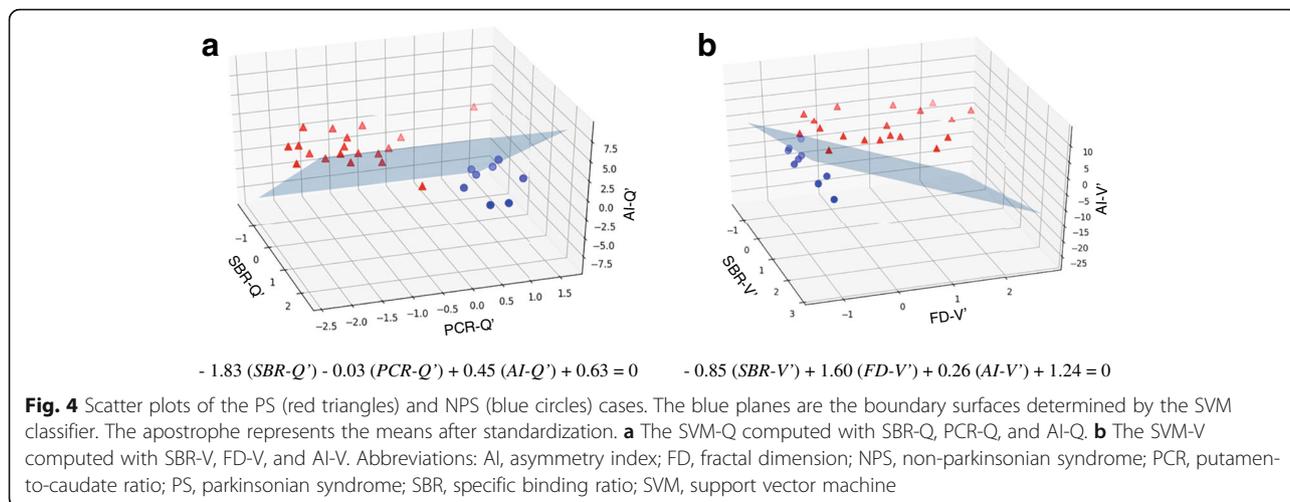
### Combined diagnostic accuracy of the three types of indices

Figure 4 shows the results of the combined diagnosis of the three types of indices using the SVM classifier. The blue planes are the boundary surface that divides the PS and NPS groups. Figure 5 shows the box-and-whisker plots and the ROC curve for SVM-Q and SVM-V. Significant differences were observed for both the SVM-Q (Fig. 5a) and SVM-V (Fig. 5b) results ( $-0.76 \pm 0.17$  vs  $0.84 \pm 0.36$ ,

**Table 2** Sensitivity, specificity, PPV, NPV, and accuracy of the six indices

	Sensitivity (%)	Specificity (%)	PPV (%)	NPV (%)	Accuracy (%)
SBR-Q	94.4 (67/71)	92.5 (37/40)	95.7 (67/70)	90.2 (37/41)	93.7 (104/111)
SBR-V	78.9 (56/71)	85.0 (34/40)	90.3 (56/62)	69.4 (34/49)	81.1 (90/111)
PCR-Q	80.3 (57/71)	87.5 (35/40)	91.9 (57/62)	71.4 (35/49)	82.9 (92/111)
FD-V	90.1 (64/71)	97.5 (39/40)	98.5 (64/65)	84.8 (39/46)	92.8 (103/111)
AI-Q	71.8 (51/71)	85.0 (34/40)	89.5 (51/57)	63.0 (34/54)	76.6 (85/111)
AI-V	73.2 (52/71)	85.0 (34/40)	89.7 (52/58)	64.2 (34/53)	77.5 (86/111)

Abbreviations: PPV positive predictive value, NPV negative predictive value, SBR specific binding ratio, PCR putamen-to-caudate ratio, FD fractal dimension, AI asymmetry index



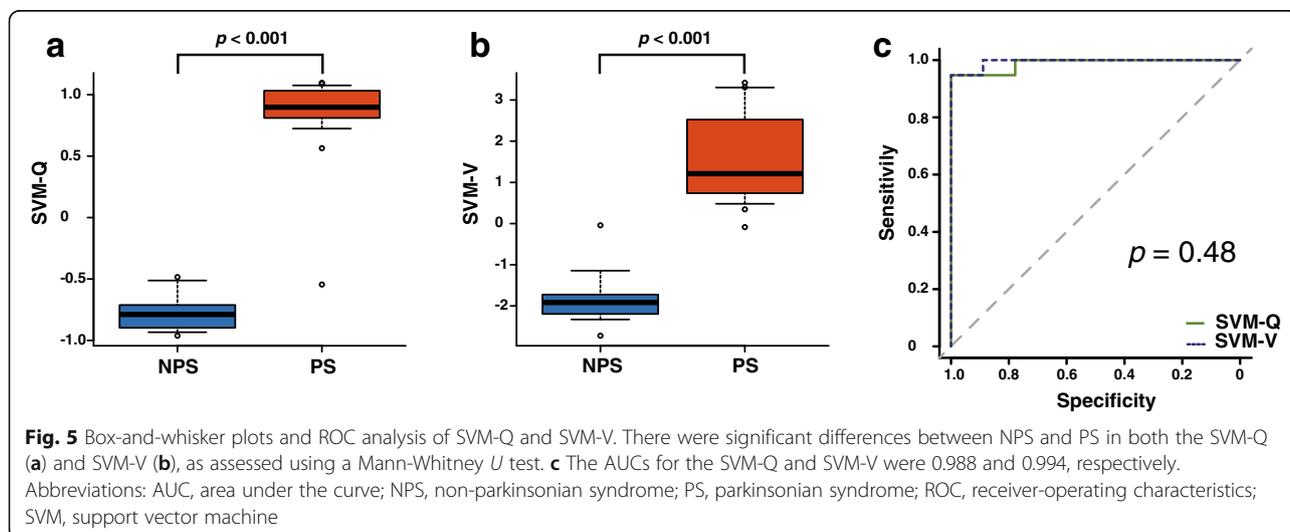
$p < 0.001$ ;  $-1.78 \pm 0.75$  vs  $1.56 \pm 1.14$ ,  $p < 0.001$ ; respectively). The AUCs for SVM-Q and SVM-V were 0.988 and 0.994, respectively (Fig. 5c). No significant difference was observed between SVM-Q and SVM-V. The sensitivity, specificity, PPV, NPV, and accuracy are summarized in Table 3.

**Discussion**

The combined diagnostic accuracy of the three types of indices (AUC for SVM-Q = 0.988, AUC for SVM-V = 0.994) was numerically higher than the individual accuracies of the six indices (AUC = 0.802–0.978). This result suggests that the combination of the three features is useful, even for quantitative evaluation, similar to the visual interpretation described in the SNM practice guideline for DAT imaging [6]. Although several studies have demonstrated the utility of each index individually [7–11], this study is the first to quantitatively evaluate the combined usefulness of these indices. We believe that our proposed method involving a

combination of the three characteristic indices can easily be used in routine clinical practice because these indices can be obtained simultaneously using simple VOI settings. This method can also reduce inter-operator variability as operators will only have to set the striatal VOI once; therefore, improvements in diagnostic accuracy may be achieved.

Our results demonstrate that no significant difference was observed when comparing the combined accuracies between two different types of VOI settings ( $p = 0.48$ ). However, the SBR-Q delivered better accuracy than the SBR-V ( $p < 0.01$ ). The method of Tossici-Bolt et al. can be influenced by the dilatation of the ventricles or the cerebral sulcus because this method uses a large VOI for the striatum [27, 28]. This might be why the accuracy of SBR-V was lower than that of SBR-Q in our study. In addition, FD-V showed better accuracy than PCR-Q ( $p < 0.001$ ). PCR-Q simply represents the ratio between the anterior and posterior striatal accumulation,



**Table 3** Sensitivity, specificity, PPV, NPV, and accuracy of the SVM-Q and SVM-V

	Sensitivity (%)	Specificity (%)	PPV (%)	NPV (%)	Accuracy (%)
SVM-Q	94.7 (18/19)	100 (9/9)	100 (18/18)	90.0 (9/10)	96.4 (27/28)
SVM-V	94.7 (18/19)	100 (9/9)	100 (18/18)	90.0 (9/10)	96.4 (27/28)

**Abbreviations:** PPV positive predictive value, NPV negative predictive value, SVM support vector machine

while FD-V can represent a change in the image through different stages using a calculation based on the threshold settings of the five stages [4]. We speculate that FD-V might be able to capture a change of distribution in more detail than PCR-Q. These results suggest that the combined use of the indices might compensate for individual differences between the corresponding indices of the two VOI settings.

To our knowledge, this is the first study to use the SVM classifier to combine the three types of indices (SBR, PCR/FD, and AI). Although several studies have evaluated the utility of SVM analysis in assessing DAT SPECT [29–32], no report has applied the three indices as features. Oliveira et al. used seven indices as features, including the SBR and PCR, as well as the length and volume of striatum uptake [29]. However, they did not include the AI index as a feature, and the seven indices used in their study cannot be obtained simultaneously using one VOI setting. Iwabuchi et al. previously assessed the combined accuracy of SBR and FD using an unsophisticated method of simply dividing SBR by FD, and did not use an SVM classifier to combine these quantitative indices, or include AI as a characteristic quantitative index [4].

This study had some limitations. First, the number of patients enrolled was relatively small. Additional studies with larger numbers of participants are needed to confirm our observations. Second, the diagnoses of PD and NPS were based on clinical observations, not pathology. Although this may have influenced our results, it is practically difficult to perform DAT SPECT and pathological examination at the same time. When autopsy is used as a reference standard, diagnostic discrepancies between DAT SPECT and autopsy findings can be noted, as the diagnosis can change with disease progression [33]. Third, we did not assess the effect of reconstruction parameters and scanner characterization. As indicated by Tossici-Bolt et al. [13], these factors would have differing effects on different VOI settings, and this matter requires further investigation.

## Conclusion

A combination of the three types of indices obtained using the SVM classifier improved the diagnostic performance for PS, and no significant difference was observed between

the combined diagnostic performance of indices calculated using the VOI settings from two different software packages.

## Abbreviations

AD: Alzheimer's disease; AI: Asymmetry index; AUC: The areas under the curve; DAT: Dopamine transporter; DLB: Dementia with Lewy bodies; FD: Fractal dimension; MSA: Multiple system atrophy; NPS: Non-parkinsonian syndrome; OSEM: Ordered-subset expectation maximization; PCR: Putamen-to-caudate ratio; PS: Parkinsonian syndrome; PSP: Progressive supranuclear palsy; RI: Radioisotope; ROC: Receiver-operating characteristics; SBR: Specific binding ratio; SPECT: Single-photon emission computed tomography; SVM: Support vector machine; VOI: Volume of interest

## Acknowledgements

The authors thank the staff of the Division of Nuclear Medicine at the Department of Radiology for their valuable support.

## Funding

This research did not receive any specific grant from funding agencies in the public, commercial, or not-for-profit sectors.

## Availability of data and materials

All data generated or analysed during this study are included in this published article.

## Authors' contributions

YI, TN, MK, and MJ were responsible for the design of the study; YI and YM prepared the SPECT data; YI and YY performed the statistical analysis; TO, DI, and HT supported the selection of clinical data; MH supported the analysis of image processing; and all authors read and critiqued the manuscript and approved the final version of the manuscript.

## Ethics approval and consent to participate

This study was approved by the Ethics Committee of Keio University School of Medicine (registration number: 20150441). This article does not contain any studies with animals performed by any of the authors.

## Consent for publication

All procedures performed in studies involving human participants were in accordance with the ethical standards of the institutional and/or national research committee and with the 1964 Helsinki declaration and its later amendments or comparable ethical standards. The institutional review board of the hospital granted permission for this retrospective review of imaging and clinical data, and waived the requirement for obtaining informed consent from the patients.

## Competing interests

TN and MJ received research grants from Nihon Medi-Physics Co., Ltd. and GE Healthcare Corp. MK received a research grant from Nihon Medi-Physics Co., Ltd. The authors declare that they have no competing interests.

## Publisher's Note

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

## Author details

<sup>1</sup>Department of Radiology, Keio University School of Medicine, 35 Shinanomachi, Shinjyuku-ku, Tokyo 160-8582, Japan. <sup>2</sup>Department of Diagnostic Radiology, Tokyo Metropolitan Geriatric Hospital and Institute of Gerontology, 35-2 Sakaecho, Itabashi-ku, Tokyo 173-0015, Japan. <sup>3</sup>Department of Neurology, Keio University School of Medicine, 35 Shinanomachi, Shinjyuku-ku, Tokyo 160-8582, Japan. <sup>4</sup>Department of Neuropsychiatry, Keio University School of Medicine, 35 Shinanomachi, Shinjyuku-ku, Tokyo 160-8582, Japan.

Received: 20 November 2018 Accepted: 18 January 2019

Published online: 28 January 2019

## References

- Varrone A, Dickson JC, Tossici-Bolt L, Sera T, Asenbaum S, Booij J, et al. European multicentre database of healthy controls for [123I]FP-CIT SPECT (ENC-DAT): age-related effects, gender differences and evaluation of different methods of analysis. *Eur J Nucl Med Mol Imaging*. 2013;40:213–27.
- Koch W, Radau PE, Hamann C, Tatsch K. Clinical testing of an optimized software solution for an automated, observer-independent evaluation of dopamine transporter SPECT studies. *J Nucl Med*. 2005;46:1109–18.
- Tossici-Bolt L, Hoffmann SM, Kemp PM, Mehta RL, Fleming JS. Quantification of [123I]FP-CIT SPECT brain images: an accurate technique for measurement of the specific binding ratio. *Eur J Nucl Med Mol Imaging*. 2006;33:1491–9.
- Iwabuchi Y, Nakahara T, Kameyama M, Yamada Y, Hashimoto M, Ogata Y, et al. Quantitative evaluation of the tracer distribution in dopamine transporter SPECT for objective interpretation. *Ann Nucl Med*. 2018;32:363–71.
- Michallek F, Dewey M. Fractal analysis in radiological and nuclear medicine perfusion imaging: a systematic review. *Eur Radiol*. 2014;24:60–9.
- Djang DS, Janssen MJ, Bohnen N, Booij J, Henderson TA, Herholz K, et al. SNM practice guideline for dopamine transporter imaging with 123I-hoflopan SPECT 1.0. *J Nucl Med*. 2012;53:154–63.
- El Fakhri G, Habert MO, Maksud P, Kas A, Malek Z, Kijewski MF, et al. Quantitative simultaneous (99m)Tc-ECD/123I-FP-CIT SPECT in Parkinson's disease and multiple system atrophy. *Eur J Nucl Med Mol Imaging*. 2006;33:87–92.
- Varrone A, Marek KL, Jennings D, Innis RB, Seibyl JP. [(123I)]beta-CIT SPECT imaging demonstrates reduced density of striatal dopamine transporters in Parkinson's disease and multiple system atrophy. *Mov Disord*. 2001;16:1023–32.
- Niimi Y, Ito S, Murate K, Hirota S, Hikichi C, Ishikawa T, et al. Usefulness of combining (123I)-FP-CIT-SPECT striatal asymmetry index and cardiac (123I)-metaiodobenzylguanidine scintigraphy examinations for diagnosis of parkinsonisms. *J Neurol Sci*. 2017;377:174–8.
- Zijlmans J, Evans A, Fontes F, Katzenschlager R, Gacinovic S, Lees AJ, et al. [123I] FP-CIT SPECT study in vascular parkinsonism and Parkinson's disease. *Mov Disord*. 2007;22:1278–85.
- Pirker W, Asenbaum S, Bencsits G, Prayer D, Gerschlag W, Deecke L, et al. [123I]beta-CIT SPECT in multiple system atrophy, progressive supranuclear palsy, and corticobasal degeneration. *Mov Disord*. 2000;15:1158–67.
- Shimizu S, Namioka N, Hirose D, Kanetaka H, Hirao K, Hatanaka H, et al. Comparison of diagnostic utility of semi-quantitative analysis for DAT-SPECT for distinguishing DLB from AD. *J Neurol Sci*. 2017;377:50–4.
- Tossici-Bolt L, Dickson JC, Sera T, Booij J, Asenbaum N, S, Bagnara MC, et al. [(123I)]FP-CIT ENC-DAT normal database: the impact of the reconstruction and quantification methods. *EJNMMI Phys*. 2017;4:8.
- Yin TK, Lee BF, Yang YK, Chiu NT. Differences of various region-of-interest methods for measuring dopamine transporter availability using 99mTc-TRODAT-1 SPECT. *Sci World J*. 2014;2014:837439.
- McKeith IG, Boeve BF, Dickson DW, Halliday G, Taylor JP, Weintraub D, et al. Diagnosis and management of dementia with Lewy bodies: fourth consensus report of the DLB consortium. *Neurology*. 2017;89:88–100.
- Gilman S, Wenning GK, Low PA, Brooks DJ, Mathias CJ, Trojanowski JQ, et al. Second consensus statement on the diagnosis of multiple system atrophy. *Neurology*. 2008;71:670–6.
- Hoglinger GU, Respondek G, Stamelou M, Kurz C, Josephs KA, Lang AE, et al. Clinical diagnosis of progressive supranuclear palsy: the movement disorder society criteria. *Mov Disord*. 2017;32:853–64.
- Nagao M, Murase K, Ichiki T, Sakai S, Yasuhara Y, Ikezoe J. Quantitative analysis of technegas SPECT: evaluation of regional severity of emphysema. *J Nucl Med*. 2000;41:590–5.
- Nagao M, Murase K, Kikuchi T, Ikeda M, Nebu A, Fukuhara R, et al. Fractal analysis of cerebral blood flow distribution in Alzheimer's disease. *J Nucl Med*. 2001;42:1446–50.
- Nagao M, Murase K, Yasuhara Y, Ikezoe J. Quantitative analysis of pulmonary emphysema: three-dimensional fractal analysis of single-photon emission computed tomography images obtained with a carbon particle radioaerosol. *AJR Am J Roentgenol*. 1998;171:1657–63.
- Nagao M, Sugawara Y, Ikeda M, Fukuhara R, Hokoishi K, Murase K, et al. Heterogeneity of cerebral blood flow in frontotemporal lobar degeneration and Alzheimer's disease. *Eur J Nucl Med Mol Imaging*. 2004;31:162–8.
- Nagao M, Sugawara Y, Ikeda M, Fukuhara R, Ishikawa T, Murase K, et al. Heterogeneity of posterior limbic perfusion in very early Alzheimer's disease. *Neurosci Res*. 2006;55:285–91.
- Vapnik V, Lerner A. Pattern recognition using generalized portrait method. *Autom Remote Control*. 1963;24:774–80.
- Vapnik VN. The nature of statistical learning theory. Berlin: Springer-Verlag; 1995.
- Kohavi R. A study of cross-validation and bootstrap for accuracy estimation and model selection. Proceedings of the 14th international joint conference on artificial intelligence - volume 2. Montreal, Quebec: Morgan Kaufmann Publishers Inc.; 1995. p. 1137–43.
- DeLong ER, DeLong DM, Clarke-Pearson DL. Comparing the areas under two or more correlated receiver operating characteristic curves: a nonparametric approach. *Biometrics*. 1988;44:837–45.
- Nonokuma M, Kuwabara Y, Hida K, Tani T, Takano K, Yoshimitsu K. Optimal ROI setting on the anatomically normalized I-123 FP-CIT images using high-resolution SPECT. *Ann Nucl Med*. 2016;30:637–44.
- Mizumura S, Nishikawa K, Murata A, Yoshimura K, Ishii N, Kokubo T, et al. Improvement in the measurement error of the specific binding ratio in dopamine transporter SPECT imaging due to exclusion of the cerebrospinal fluid fraction using the threshold of voxel RI count. *Ann Nucl Med*. 2018;32:288–96.
- Oliveira FPM, Faria DB, Costa DC, Castelo-Branco M, Tavares J. Extraction, selection and comparison of features for an effective automated computer-aided diagnosis of Parkinson's disease based on [(123I)]FP-CIT SPECT images. *Eur J Nucl Med Mol Imaging*. 2017;45:1052–62.
- Illan IA, Gorz JM, Ramirez J, Segovia F, Jimenez-Hoyuela JM, Ortega Lozano SJ. Automatic assistance to Parkinson's disease diagnosis in DaTSCAN SPECT imaging. *Med Phys*. 2012;39:5971–80.
- Huertas-Fernandez I, Garcia-Gomez FJ, Garcia-Solis D, Benitez-Rivero S, Marin-Oyaga VA, Jesus S, et al. Machine learning models for the differential diagnosis of vascular parkinsonism and Parkinson's disease using [(123I)]FP-CIT SPECT. *Eur J Nucl Med Mol Imaging*. 2015;42:112–9.
- Taylor JC, Fenner JW. Comparison of machine learning and semi-quantification algorithms for (123I)FP-CIT classification: the beginning of the end for semi-quantification? *EJNMMI Physics*. 2017;4:29.
- Grandal Leiros B, Perez Mendez LI, Zelaya Huerta MV, Moreno Eguinoa L, Garcia-Bragado F, Tunon Alvarez T, et al. Prevalence and concordance between the clinical and the post-mortem diagnosis of dementia in a psychogeriatric clinic. *Neurologia*. 2018;33:13–7.

Submit your manuscript to a SpringerOpen® journal and benefit from:

- Convenient online submission
- Rigorous peer review
- Open access: articles freely available online
- High visibility within the field
- Retaining the copyright to your article

Submit your next manuscript at ► [springeropen.com](https://www.springeropen.com)