# Finding the optimal cutoff value for amyloid β positivity using the iterative outlier method and concordance rate

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#### ABSTRACT

**Purpose:** The purpose of this study was to calculate the cutoff value for amyloid  $\beta$  (A $\beta$ ) positron emission tomography (PET) positivity using the iterative outlier method and to evaluate its validity based on the concordance rate.

**Methods:** We performed the iterative outlier method on 373 cognitively unimpaired (CU) subjects and calculated the optimal cutoff value for A $\beta$  positivity. The validation was performed using the independent dataset, comprising 83 subjects (27 CU, 27 amnestic mild cognitive impairment, and 29 Alzheimer's dementia). We evaluated the validity of the A $\beta$  cutoff value by calculating its concordance rate with the visual assessment and between two different A $\beta$  tracers performed in the same subject.

**Results:** The concordance rate of A $\beta$  cutoff values with the visual assessment ranged from 84.3% to 91.5%, depending on the reference regions. The concordance rate of the cutoff values between the two A $\beta$  tracers ranged from 90.3% to 97.5%, all of which were higher than that of the visual assessment (86.7%).

**Conclusion:** We demonstrated that the iterative outlier method could identify the cutoff value for  $A\beta$  PET positivity.

Keywords: Alzheimer disease; Amyloid; Positron-emission tomography

Optimal cutoff value for amyloid  $\beta$  positivity

#### **INTRODUCTION**

Alzheimer's disease (AD) is the most common cause of dementia in the elderly [1], and is characterized by the deposition of amyloid  $\beta$  (A $\beta$ ) in the brain. The amyloid cascade hypothesis, a well-known hypothetical model of AD, posits that accumulation of A $\beta$  is the earliest pathogenic process leading to tau deposition, neurodegeneration, and clinical cognitive impairment [2]. Therefore, distinguishing brains free of A $\beta$ from those with A $\beta$  deposition is of utmost importance for the early diagnosis of AD [3].

A $\beta$  deposition in the brain can be measured directly by brain autopsy or indirectly by either cerebrospinal fluid (CSF) sampling or A $\beta$  tracing positron emission tomography (PET) [4]. A $\beta$  PET is more advantageous than CSF sampling as it is less invasive and detects regional A $\beta$  depositions. Measurement of A $\beta$  deposition by PET is by either visual assessment or more objective measures, such as by calculating the standard uptake value ratio (SUVR) [5] or distribution volume ratio (DVR) [6]. However, since both SUVR and DVR are continuous measures, it is necessary to dichotomize them into evidence (positive) or no evidence of deposition (negative).

Previous studies have presented a variety of approaches to define A $\beta$ -positive cutoffs, including the iterative outlier method [7]. Until now, the validity of these cutoffs has often been evaluated with the standard-of-truth (SOT) based on either histological findings of brain autopsy or visual assessment of PET images. However, visual assessment is operator-dependent [8] and it is difficult to obtain postmortem data.

In this study, we calculated the cutoff value for A $\beta$  positivity using the iterative outlier method and evaluated its validity based on the concordance rate. Two tracers were used to evaluate the concordance rate of A $\beta$  positivity based on the cutoff value in the same subject.

#### **METHODS**

#### **Development dataset**

For the iterative outlier method, we used A $\beta$  PET data obtained from 373 cognitively unimpaired (CU) subjects aged over 50 years. CU subjects were defined as those with normal results on neurological examination and normal cognitive function, expected for their ages and education, on a standardized cognitive test (Seoul Neuropsychological Screening Battery) [9]. Of the 373 subjects, 202 and 171 subjects underwent <sup>18</sup>F-florbetaben (FBB) and <sup>18</sup>F-flutemetamol (FMM) A $\beta$  PET, respectively.

#### Validation dataset

For validation of the A $\beta$ -positive cutoff value, we used the head-to-head comparison data, comprising 83 subjects (27 CU, 27 amnestic mild cognitive impairment, and 29 AD), in whom both FBB and FMM A $\beta$  PET were performed. A detailed description is provided in our previous study [10].

#### Magnetic resonance imaging acquisition

Standardized three-dimensional T1 turbo field echo images were acquired from all subjects at the Samsung Medical Center using the same 3.0-T magnetic resonance imaging (MRI) scanner (Philips Achieva, Philips Healthcare, Andover, MA, USA). The following parameters were used: sagittal slice thickness of 1.0 mm, over contiguous slices with 50% overlap, no gap, repetition time of 9.9 ms, echo time of 4.6 ms, flip angle of 8°, and matrix size of  $240 \times 240$  pixels, reconstructed to  $480 \times 480$  over a field of view of 240 mm.

#### **PET** acquisition

PET images were acquired using a Discovery STe PET/computed tomography (CT) scanner (GE Medical Systems, Milwaukee, WI, USA) in three-dimensional scanning mode that examined 47 slices of 3.3-mm thickness spanning the entire brain. CT images were acquired using a 16-slice helical CT (140 KeV, 80 mA; 3.75-mm section width) for attenuation correction. According to protocols proposed by the ligand manufacturers, a 20-minute emission PET scan with dynamic mode (consisting of  $4 \times 5$  minutes frames) was performed 90 minutes after injection of a mean dose of 311.5 MBg FBB and 185 MBq FMM, respectively. Three-dimensional PET images were reconstructed in a  $128 \times 128 \times 48$  matrix with a voxel size of  $2 \times 2 \times 3.27$  mm using the ordered-subsets expectation maximization algorithm (FBB iterations = 4 and subset = 20; FMM iterations = 4 and subset = 20). AB positivity was determined based on visual assessments performed by three experienced raters (two nuclear medicine doctors and one neurologist) [11-15].

#### Image preprocessing and calculation of the SUVR

To calculate the SUVR, we performed the following processes. First, we co-registered the PET image to the T1-MRI images. Subsequently, we normalized the native PET image to the Montreal Neurological Institute-152 template using the transformation matrix, calculated during the segmentation of T1-MRI. After normalization, the brain was divided into 116 grey

Hang-Rai Kim, et al.

matter regions using the automated anatomical labeling (AAL) atlas [16]. We considered three reference regions (whole cerebellum [WC], cerebellar gray matter [CG], and pons) to calculate the SUVR, the regional masks of which were obtained from the GAAIN website (http://www.GAAIN.org) [17]. We calculated the global cortical SUVR as the volume-weighted mean of 56 cortical regions in the AAL template [18]. This pre-processing was performed using SPM8 through Matlab 2014b (Mathworks, Natick, MA, USA).

#### Iterative outlier method

We used the iterative outlier method and generated an upper- and lower-bound SUVR. During the iteration, subjects with greater than the upper inner-bound (3rd quartile+1.5 interquartile range [IQR] or less than the lower inner-bound [1st quartile-1.5 IQR]) values were removed from the dataset. This process was repeated until all outliers were removed. In the final dataset, a cutoff value was determined by adding 2.5% of the maximum SUVR in itself [19].

#### Statistical analysis

We performed two analyses using the independent dataset to validate the A $\beta$  cutoff value. We first compared the cutoff value-derived A $\beta$  positivity with the visually determined A $\beta$ positivity and calculated the concordance rate. Next, we evaluated whether the cutoff value-derived A $\beta$  positivity of one subject was concordant between the two A $\beta$  tracers (FBB and FMM). We assumed that the optimal cutoff value would distinguish subjects with positive A $\beta$  deposition regardless of the tracer type. The statistical significance of the concordance rate was evaluated using the McNemar's chisquare test, with a P-value of less than 0.05 indicating non-

Table 1.	Demograp	hics of the	e develop	ment and	validation	datasets

agreement between the two measures. All statistical analyses were completed using R version 3.5.3 (http://www.r-project.org/). This is a retrospective study using the data from the preceding study [10], which was approved by the Institutional Review Board of Samsung Medical Center (2015-04-091) with written informed consent of the participants at the time of enrollment.

#### RESULTS

#### Subject demographics

Table 1 shows the demographics of the development and validation datasets. In the development dataset, subjects who underwent FBB were older than those who underwent FMM. In the validation dataset, although two different A $\beta$  PETs were performed in the same subject, the frequency of A  $\beta$  positivity differed between the two tracers.

#### Aß cutoff value with the iterative outlier method

Using the iterative outlier method, we calculated the optimal cutoff value for A $\beta$  PET positivity according to the A $\beta$  tracers and reference regions (Fig. 1). In both FBB and FMM, the cutoff value was higher when the CG was used as the reference region.

### Validation of the cutoff value based on the concordance rate with the visual assessment

To validate the cutoff value derived from the iterative outlier method, we applied the cutoff value for A $\beta$  positivity in the independent dataset and compared it with that of the visual assessment (Table 2). The concordance rate ranged from 84.3% to 91.5%, and the highest rate was observed when the

Domographic	Developm	ent dataset	Validation dataset		
Demographic –	FBB (n=202)	FMM (n=171)	FBB (n=83)	FMM (n=83)	
Age (yr)	71.4±7.2	68.7±7.4	71.2±7.7		
Women	118 (58.4)	109 (63.7)	50 (60.2)		
Education (yr)	12.6±4.4	11.6±5.0	9.8±4.8		
MMSE, score	28.3±1.5	28.0±2.1	25.5	±5.0	
Diagnosis (CU/aMCI/AD)	202/0/0	171/0/0	27/2	7/29	
Aβ positivity <sup>a)</sup>	32 (15.8)	27 (15.8)	45 (54.2)	52 (62.7)	

Values are presented as mean±standard deviation or number (%).

FBB, <sup>18</sup>F-florbetaben; FMM, <sup>18</sup>F-flutemetamol; MMSE, Mini-Mental State Examination; CU, cognitive unimpaired; aMCI, amnestic mild cognitive impairment; AD, Alzheimer's disease; Aβ, amyloid β.

<sup>a)</sup>Aβ positivity was determined based on visual assessments.

Optimal cutoff value for amyloid ß positivity



Fig. 1. Results of the iterative outlier method for (A, B, C) <sup>18</sup>F-florbetaben (FBB) and (D, E, F) <sup>18</sup>F-flutemetamol (FMM). Red lines indicate cutoff value derived using the iterative outlier method. SUVR, standard uptake value ratio; CG, cerebellar gray matter; WC, whole cerebellum.

Aβ tracer	Reference region	Concordance rate	P-value <sup>a)</sup>
FBB	CG	0.855	0.773
	WC	0.915	0.450
	Pons	0.903	0.724
FMM	CG	0.843	0.579
	WC	0.855	0.773
	Pons	0.843	0.579

 Table 2. The concordance rate of the cutoff value with visual assessment

A $\beta$ , amyloid  $\beta$ ; FBB, <sup>18</sup>F-florbetaben; CG, cerebellar gray matter; WC, whole cerebellum; FMM, <sup>18</sup>F-flutemetamol.

<sup>a)</sup>P-value was calculated using McNemar's chi-square test.

WC was used as the reference for both FBB and FMM. All P-values were greater than 0.05, indicating that the two measures (the cutoff value and the visual assessment) were in accordance.

#### Validation of the cutoff value based on the concordance rate between FBB and FMM We evaluated whether the cutoff value-derived Aβ positivity

Table 3. The concordance rate of the SUVR cutoff between the two  $A\beta$  tracers

Methods	Concordance rate	P-value <sup>a)</sup>
SUVR cutoff		
CG	0.903	0.077
WC	0.975	0.479
Pons	0.975	0.479
Visual assessment	0.867	0.070

SUVR, standard uptake value ratio; A $\beta$ , amyloid  $\beta$ ; CG, cerebellar gray matter; WC, whole cerebellum.

<sup>a)</sup>P-value was calculated using McNemar's chi-square test.

was consistent between the two tracers. The concordance rate ranged from 90.3% to 97.5%, all of which were higher than that of the visual assessment (86.7%).

The concordance rate of visual assessment between FBB and FMM was 86.7% (72/83) (Table 3, Fig. 2). There were 11 subjects who showed discordant results (nine subjects [FBB negative/FMM positive] and two subjects [FBB positive/FMM negative]). The concordance rates of the cutoff values were 90.3% (75/83), 97.5% (81/83), and 97.5% (81/83) for each reference region (CG, WC, and pons, respectively) (Table 3). The

Hang-Rai Kim, et al.



Fig. 2. Scatter plot of <sup>18</sup>F-florbetaben (FBB) and <sup>18</sup>F-flutemetamol (FMM) standard uptake value ratio (SUVR) with (A) cerebellar gray matter, (B) whole cerebellum, and (C) pons as the reference regions. Dotted lines indicate the cutoff value.

majority of the subjects who showed discordant results in the visual assessment showed concordant results when the cutoff value was used (9/11 for the CG [Fig. 2A], 11/11 for the WC [Fig. 2B], and 10/11 for the pons [Fig. 2C]).

#### DISCUSSION

In this study, we identified the optimal SUVR cutoff for  $A\beta$  positivity using the iterative outlier method and evaluated its validity based on the concordance rate.

We found that the cutoff optimization methods were comparable to visual assessment. Detecting individuals with Aß deposition is important, as it is valuable for the clinical diagnosis of AD and in clinical trials aiming to reduce the burden of AB in the brain [4]. To date, in the clinical field, AB positivity of PET is generally determined by experienced clinicians based on visual rating. However, visual rating can be operator- and AB tracer-dependent, thus are prone to intra- and inter-rater discrepancies [8]. Interestingly, we found that the iterative outlier method-derived cutoff value had a higher concordance rate of AB positivity between the two tracers than that of the visual assessment. Although its superiority to visual assessment should be tested statistically, this finding suggests that the iterative outlier method-derived cutoff value could be an accurate measure of  $A\beta$  positivity in FBB and FMM PET.

Although the iterative outlier method may have an advantage over visual assessment, there are other approaches to define  $A\beta$ -positive cutoff values, such as receiver operating characteristic (ROC) analyses [20] and clustering methods [21]. The important difference in the approaches is that the iterative outlier method requires an annotated group of cognitively healthy subjects [7], whereas the ROC method requires both diseased and healthy groups [20], and the cluster analysis requires mixed groups that do not require annotations [21].

We further observed that cutoff optimization methods differed according to the reference regions and types of A $\beta$  tracers. This discrepancy precluded the generalization of findings from one A $\beta$  tracer to another [22]. Therefore, studies based on A $\beta$  PET data should describe how the data is processed along with a description of the reference regions and the type of tracer.

The strengths of this study are that we used a data-driven approach to identify cutoff values and used a unique cohort (head-to-head comparison data) for the validation, where two different types of A $\beta$  PET were performed in the same subject. However, this study has several limitations. First, we did not perform a statistical test to compare newly derived cutoff values with results of the visual assessment. This can be done by the bootstrapping method; however, the small sample size of our study precluded this analysis. Second, the iterative outlier method depends on the dataset used. Although we tried to obtain a homogenous dataset by thorough examinations, it is possible that the cutoff value will change if more subjects are included. Therefore, our findings should be interpreted with caution and replicated using a larger dataset. Lastly, ideally, histopathological confirmation

Optimal cutoff value for amyloid β positivity

of A $\beta$  deposition in the brain should be the SOT. Due to the scarcity of post-mortem data, we considered visual assessment of A $\beta$  PET as the SOT in this study.

In conclusion, we demonstrated that the iterative outlier method could identify the cutoff value for A $\beta$  positivity in both FBB and FMM PET.

#### **CONFLICTS OF INTEREST**

No potential conflict of interest relevant to this article was reported. Sang Won Seo has been editorial board of P&FM since December 2017. He was not involved in the review process of this article.

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Acquisition, analysis, or interpretation of data: HRK, YSC, SHM, HJK, HJ, DLN, SP, SWS.

Drafting the work or revision: HRK, YSC, SWS.

Final approval of the manuscript: HRK, YSC, SWS.

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Hang-Rai Kim, et al.

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