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Derivation and utility of an A β -PET pathology accumulation index to estimate A β load

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Abstract

Objective: To evaluate a novel A β -PET based quantitative measure (A β accumulation index [A β -index]), including the assessment of its ability to discriminate between subjects based on A β -status using visual-read, CSF A β 42/A β 40 and *post-mortem* neuritic-plaque burden as standards of truth.

Methods: 1121 subjects (with and without cognitive impairment) scanned with A β -PET: Swedish BioFINDER, n=392, [18 F]flutemetamol; ADNI, n=692, [18 F]florbetapir; a phase-3 end-of-life study, n=100, [18 F]flutemetamol). The relationships between A β -index and standardized uptake values ratios (SUVR) from A β -PET were assessed. The diagnostic performance of A β -index and SUVR were compared when using visual reads, CSF A β 42/A β 40 and A β -histopathology as reference standards.

Results: Strong associations were observed between A β -index and SUVR (R^2 , BioFINDER, 0.951; ADNI, 0.943, end-of-life, 0.916). Both measures performed equally well in differentiating A β -positive from A β -negative subjects, with AUCs of 0.979-0.991 to detect abnormal visual reads, AUCs of 0.961-0.966 to detect abnormal CSF A β 42/40 and AUCs of 0.820-0.823 to detect abnormal A β -histopathology. Both measures also showed a similar distribution across *post-mortem* based A β -phases (based on anti-A β 4G8 antibodies). By comparison to models using visual-read alone, the addition of the A β -index resulted in a significant increase in AUC and a decrease in Akaike information criterion to detect abnormal A β -histopathology.

Conclusions: The proposed A β -index showed a tight association to SUVR and carries an advantage over the latter in that it does not require the definition of regions of interest nor the use of MRI. A β -index may thus prove simpler to implement in clinical settings and may also facilitate the comparison of findings using different A β -PET tracers.

Classification of evidence

This study provides Class III evidence that the A β accumulation index accurately differentiates A β -positive from A β -negative subjects when compared to A β -PET visual reads, CSF A β 42/A β 40 and A β -histopathology.

Introduction

A β -PET tracers such as [¹⁸F]flutemetamol are currently approved for visual assessment only (whereby images are rated as negative/positive (normal/abnormal) by a trained rater).^{1,2}

Evidence suggests, however, that quantifying the amount of tracer retention in the brain may improve agreement between raters³ and aid in the monitoring of treatment effects in anti-A β trials.⁴ The most commonly used quantitative measure for A β -PET is the standardized uptake value ratio (SUVR), where tracer concentration in cortical (target) regions is divided by that within a reference region assumed to be free of A β -pathology. Though high resolution structural magnetic resonance imaging (MRI) is ideally used to delineate these regions-of-interest (ROIs), this technique is not always available in clinical settings and is frequently contraindicated among elderly patient.⁵

In the absence of MRI, PET-only approaches can be used.⁶ Here, a PET image is first transformed from a standardized coordinate space (spatial normalization); ROIs are then defined using a probabilistic atlas. Using such an approach, we recently described a novel measure of brain A β -burden (A β -index).⁷ As it does not require the definition of ROIs, it is simpler to implement than SUVR. Herein, we aimed to 1) compare A β -index and SUVR in three independent cohorts; 2) to assess their ability to differentiate A β -positive and A β -negative subjects using visual read, CSF A β 42/A β 40 or *post-mortem* neuritic plaque burden as standards-of-truth.; and 3) to assess whether a combination of visual read and A β -index

was superior to visual read alone to predict A β -positivity using CSF A β 42/A β 40 and *post-mortem* neuritic plaque burden as standards-of-truth. We hypothesized that across these aims, A β -index would show non-inferiority to SUVR.

Methods

Participants

Our population consisted of 1121 subjects with A β -PET, from three separate cohorts: 392 from the Swedish BioFINDER study (clinical trial no. NCT01208675), scanned with [18 F]flutemetamol (251 cognitively unimpaired (CU), including 129 elderly controls and 122 with subjective cognitive decline, and 141 CI subjects with mild cognitive impairment [MCI]), 629 from ADNI (clinical trial no. NCT00106899) (246 CU controls and 383 CI subjects [MCI]), scanned with [18 F]florbetapir, and 100 subjects from a phase 3 end-of-life study (clinical trial no. NCT01165554 and NCT02090855) who were scanned with [18 F]flutemetamol *ante-mortem* and autopsied after death.⁸⁻¹⁰ Inclusion for CU and CI individuals from BioFINDER and ADNI have been described elsewhere^{11, 12} and are described in the Supplement (Appendices e-1 and e-2, doi:10.5061/dryad.2547d7wnf). In the end-of-life study, subjects were ≥ 55 years of age, terminally ill with a life expectancy < 3 years and with general health sufficient to allow for completion of study procedures. Dementia, defined according to the DSM-IV criteria, was noted as present or absent, as reported in case notes. The relationship between A β -index and SUVR was also examined in patients with AD dementia, using CSF A β 42/A β 40 as the standard of truth for A β status (BioFINDER, n=25 with [18 F]flutemetamol PET available and n=25 from ADNI using [18 F]florbetapir).

Standard Protocol Approvals, Registrations, and Patient Consents

Written informed consent was obtained from all patients (or guardians of patients) participating in the study (consent for research). Ethical approval for BioFINDER was given by the Regional Ethical Committee of Lund University. Approval for PET imaging was obtained from the Swedish Medical Products Agency and the local Radiation Safety Committee at Skåne University Hospital. For the ADNI and end-of-life cohorts, study protocols were approved by local ethical committees.

Image Acquisition and Processing

For BioFINDER, [¹⁸F]flutemetamol studies were performed using a Philips Gemini TF PET/CT scanner (Philips Medical Systems, Amsterdam, Netherlands) over the interval 90- to 110-min post-injection; data was acquired in list-mode and binned into frames using an iterative Vue Point HD algorithm (six subsets, 18 iterations with 3mm filter and no time-of-flight correction). All participants underwent 3T MRI scans (Siemens MAGNETOM Prisma), acquiring isometric 1 mm³ T1-weighted magnetization-prepared rapid gradient echo (MPRAGE) images. For ADNI, [¹⁸F]florbetapir image data acquired 50- to 70-min post-injection¹³ and T1-weighted MRI scans using a sagittal volumetric MPRAGE sequence acquired at 3T were used.¹⁴ For the phase 3 end-of-life study, [¹⁸F]flutemetamol PET images were acquired on PET/CT cameras over the interval 90- to 110-min post-injection. As both ADNI and the phase 3 end-of-life study were multicentric in nature, images were smoothed to achieve a uniform imaging resolution.

A β -PET images were spatially normalized using two approaches: an MRI driven approach included in SPM12, and a PET driven principal component approach.⁷ For the end-of-life cohort, normalization of [¹⁸F]flutemetamol images was only performed using the principal component approach as MRI was not available. The purpose of this dual approach

was to ensure that SUVRs from the principal component approach were highly correlated with those derived using the MRI driven gold standard approach. The complete details of the principal component approach can be found in the original publication.⁷ Briefly, tracer specific principal component images are first calculated by singular value decomposition of A β -PET SUVR images. Two principal components were chosen as these captured ~95% of the variance in the dataset. The first principal component represents the average of all the images the dataset, the second, the different between A β -positive and A β -negative images (i.e. specific binding). A synthetic template ($I_{Synthetic}$) can then be modelled as a linear combination of the first and second principal component images (I_{PC1} and I_{PC2} , respectively). As part of this operation, I_{PC2} is multiplied by a weighting factor (A β -index; i.e. $I_{Synthetic} = I_{PC1} + A\beta\text{-index} * I_{PC2}$) representing a global measure of brain A β pathology. Here, a positive A β -index yields a template with a more A β -positive appearance and a negative value yields a template with a more A β -negative appearance. Using an algorithm that incorporates both the A β -index and the parameters required for the spatial transformation, the synthetic template can then be used to normalize A β -PET images.

In the present study, pre-existing synthetic templates derived from phase 2 studies were used for [¹⁸F]flutemetamol¹⁵ and [¹⁸F]florbetapir.^{16, 17} Following spatial normalization of [¹⁸F]flutemetamol and [¹⁸F]florbetapir scans, two sets of SUVRs were calculated (one for each normalization method) for BioFINDER and ADNI subjects using a composite cortical region of interest (ROI)—encompassing brain regions typically showing high A β load in AD, including frontal, temporal and parietal cortices, precuneus, anterior striatum, and insular cortex—and the pons and cerebellum as reference tissues¹⁸ for [¹⁸F]flutemetamol and [¹⁸F]florbetapir, respectively. As described above, SUVR values for [¹⁸F]flutemetamol scans from the end-of-life cohort were based on the principal component based normalization alone.

The term “principal component” herein refers to the principal component driven normalization approach (and, for instance, SUVR values derived from images normalized using this approach) and not the A β -index per se (which is generated when using the principal component driven approach for spatial normalization).

CSF Biomarkers

Lumbar puncture and CSF handling followed structured protocols in the BioFINDER and ADNI studies.¹¹ In BioFINDER, the fully automated Elecsys assays (Roche Diagnostics) were used, as described elsewhere,^{19, 20} with samples analyzed at the Clinical Neurochemistry Laboratory, University of Gothenburg, Sweden. In ADNI, CSF A β 42/A β 40 was determined using A β 42 and A β 40 measurements derived from a liquid chromatography with tandem mass spectrometric detection (UPLC-MS-MS) method,²¹ with samples analyzed in the Biomarker Research Laboratory, University of Pennsylvania, USA.

Post-Mortem A β -Pathology Assessment

Post-mortem based estimates of A β plaque pathology were based on autopsy brain tissue previously collected in support of the GE067-007/GE-067-026 phase 3 clinical trials for [¹⁸F]flutemetamol PET.²² As described elsewhere,¹⁰ after formalin fixation, brains were coronally sliced and macroscopically screened. The Bielschowsky silver method was then applied to paraffin-embedded tissue from eight neocortical regions, as defined by the Consortium to Establish a Registry for Alzheimer’s disease (CERAD).²³ In each region, neuritic plaque densities were recorded as 0 (no plaques), 1 (sparse; 1–5 plaques), 2 (moderate; 6–19 plaques), or 3 (frequent; >20 plaques), per 100x field of view.^{23, 24} A modified CERAD based assessment approach was then applied whereby the arithmetic mean of neuritic plaque density was calculated across the eight investigated regions (30 measures

per region), giving a continuous variable. The A β -phases according to Thal et al.⁹ describing the hierarchical spreading of A β plaque pathology in the brain were determined after screening the A β -stained sections (anti-A β 4G8 antibodies) for plaque distribution.

Definition of A β Status

A β status (positive/negative), as a standard of truth, was defined using three approaches: in ADNI and BioFINDER, using consensus read of A β -PET uptake images and CSF A β 42/A β 40; and in the end-of-life trial, using the *post-mortem* based Bielschowsky histopathology score. For CSF A β 42/A β 40, cut-offs of 0.059 (BioFINDER)²⁵ and 0.137 (ADNI) were used, based on Gaussian mixture modelling applied to the BioFINDER and ADNI cohorts. For Bielschowsky silver stain, each assessed brain region was scored from 0 to 3 (calculated as the arithmetic mean of 30 measures); a score above 1.5, previously shown to represent the threshold between the categories of sparse and moderate neuritic plaques,¹⁰ in any of the eight investigated regions was considered abnormal, with the brain classified as A β -positive.

Statistical Analyses

Between group characteristics were compared using Kruskal-Wallis' or Fisher's exact tests. The relationship between A β -index and SUVR from [¹⁸F]flutemetamol and [¹⁸F]florbetapir was examined using linear regression and coefficient of determination (R²). Receiver operating characteristic (ROC) analyses were performed to generate area-under-the-curve (AUC) values for both A β -index and SUVR. Differences in AUC values for these measures were evaluated using bootstrap (n=1000) procedures. In order to assess the added clinical value of the A β -index, AUC and Akaike information criterion (AIC) values from binary logistic regression models (using CSF A β 42/A β 40 (BioFINDER and ADNI) and

Bielschowsky histopathology (end-of-life cohort) as the outcome variables [positive/negative]) and visual read (model 1) or visual read in combination with A β -index (continuous, model 2) were compared. For comparison, a third model combining visual read and SUVR was included. All analyses were performed in R (v.3.5.3; <https://www.R-project.org/>), with significance set at $P < 0.05$, two-sided.

Primary research question

How do A β -index and SUVR from A β PET compare in their ability to differentiate subjects based on their A β -status. This study provides Class III evidence that the A β accumulation index accurately differentiates A β -positive from A β -negative subjects when compared to A β -PET visual reads, CSF A β 42/A β 40 and A β -histopathology.

Data Availability

Anonymized study data for the primary analyses presented in this report are available on request from any qualified investigator for purposes of replicating the results.

Results

Cohort characteristics are summarized in tables 1 (BioFINDER and ADNI) and 2 (end-of-life study). Among the 85 cases in the end-of-life cohort with an antemortem diagnosis of dementia, 28 (33%) had a post-mortem neuropathological diagnosis of pure AD, 33 (39%) a diagnosis of AD plus at least one other pathology (e.g. cerebral amyloid angiopathy or TDP-43) and 24 (28%) a non-AD pathology such as Lewy body or vascular dementia. Figure 1 provides an overview of the steps required to generate the A β -index. Principal component images for [18 F]flutemetamol and [18 F]florbetapir are provided in figure e-1 (doi:10.5061/dryad.2547d7wnf). Comparison of SUVR values using principal component and MRI (SPM12) driven normalization approaches showed good agreement (figure e-2;

doi:10.5061/dryad.2547d7wnf), with R^2 values of 0.997 for [^{18}F]flutemetamol and 0.995 for [^{18}F]florbetapir ($P < 0.001$). Characteristics of AD dementia subjects are summarized in table e-1 (doi:10.5061/dryad.2547d7wnf).

Strong associations were observed between $\text{A}\beta$ -index and principal component derived SUVR values in both cohorts (BioFINDER, $R^2 = 0.951$ [95% confidence interval, 0.933-0.961]; ADNI, $R^2 = 0.943$ [95% confidence interval, 0.927-0.952]; $P < 0.001$) (figure 2). Comparison of ROC derived AUC values from $\text{A}\beta$ -index and SUVR showed that both measures performed equally well in differentiating $\text{A}\beta$ -positive from $\text{A}\beta$ -negative subjects, using both visual read (AUCs = 0.979 [95% confidence interval, 0.972-0.989] to 0.991 [95% confidence interval, 0.972-0.989]) and CSF $\text{A}\beta_{42}/\text{A}\beta_{40}$ (AUCs = 0.961 [95% confidence interval, 0.939-0.983] to 0.971 [95% confidence interval, 0.949-0.981]) as standards of truth (figure 3), with no significant difference found between AUC values. Similar findings were obtained when looking at AD dementia cases (figure e-3; doi:10.5061/dryad.2547d7wnf).

In the [^{18}F]flutemetamol end-of-life cohort we found that both $\text{A}\beta$ -index and SUVR could predict i) abnormal Bielschowsky silver stain scores (AUCs = 0.820 [95% confidence interval, 0.716-0.923] to 0.823 [95% confidence interval, 0.725-0.921]) and ii) visual read outcomes (AUCs = 0.938 [95% confidence interval, 0.889-0.984] to 0.949 [95% confidence interval, 0.911-0.988]) (figure 4), with no significant difference found between AUC values. Using the $\text{A}\beta$ phases (Thal et al.²⁶) as neuropathological readout, a similar distribution was seen for [^{18}F]flutemetamol $\text{A}\beta$ -index and SUVR across $\text{A}\beta$ phases (figure 4). Comparison of SUVR and $\text{A}\beta$ -index measures between $\text{A}\beta$ phases (i.e. phase 1 vs 2, 2 vs 3, 3 vs 4 and 4 vs 5) showed significant differences for both measures between phases for the contrasts phase 3 vs 4 (SUVR and $\text{A}\beta$ -index, $P < 0.001$) and phase 4 vs 5 (SUVR and $\text{A}\beta$ -index, $P < 0.01$).

Finally, we studied whether a combination of A β -index and visual read was superior to visual read alone. By comparison to the binary logistic regression model using only visual read as a predictor, the addition of the A β -index resulted in a significant increase in AUC (A β -status as outcome) using CSF A β 42/A β 40 (BioFINDER: 0.868 [95% confidence interval 0.843 to 0.894] vs 0.962 [95% confidence interval 0.932-0.987], $P < 0.001$; ADNI: 0.881 [95% confidence interval 0.856-0.906] vs 0.943 [95% confidence interval 0.923-0.962], $P < 0.001$) and Bielschowsky histopathology (0.910 [95% confidence interval 0.883 to 0.937] vs 0.961 [95% confidence interval 0.936-0.986], $P < 0.05$). Moreover, addition of the A β -index resulted in an improved model fit (AIC) (using CSF A β 42/A β 40 BioFINDER: 253.94 vs 167.78; ADNI: 400.59 vs 328.89; using Bielschowsky histopathology in the end-of-life cohort, 60.24 vs 53.86).²⁷ Similar findings were observed when using SUVR for both AUC (AUC: CSF A β 42/A β 40, BioFINDER 0.868 [95% confidence interval 0.843-0.894] vs 0.942 [95% confidence interval 0.914-0.968], $P < 0.001$; ADNI: 0.881 [95% confidence interval 0.859-0.905] vs 0.954 [95% confidence interval 0.925-0.983], $P < 0.001$; Bielschowsky, end-of-life cohort, 0.910 [95% confidence interval 0.889-0.933] vs 0.942 [95% confidence interval 0.917-0.969], $P < 0.05$ and AIC (CSF A β 42/A β 40, BioFINDER: 253.94 vs 161.08; ADNI: 400.59 vs 306.31; and in the end-of-life cohort using Bielschowsky histopathology: 60.24 vs 56.94).

Discussion

The objective of the present study was to assess the relationship between A β -index and A β -PET SUVR, including a comparison of the ability of both measures to differentiate between subjects on the basis of their A β status, using several standards of truth (visual read, CSF A β 42/A β 40 and *post-mortem* A β histopathology). First, using both [¹⁸F]flutemetamol and [¹⁸F]florbetapir, we showed that the principal component based approach to normalization

was precise and accurate, with SUVRs from this approach correlating highly with those derived from the MRI based method in SPM. Using this PET driven approach, we then showed a close correspondence between the A β -index and SUVR, with both measures performing equally well in identifying A β -positive cases and showing a similar pattern of increase across *post-mortem* A β -phases. Finally, we showed that the addition of the A β -index improved prediction of A β -status relative to the use of visual read alone.

Given recent evidence showing that A β -PET imaging led to changes in the clinical management of CI individuals,²⁸ the importance of accurate and reproducible A β image interpretation is clear. Though the visual assessment of A β scans as positive or negative has been shown to be an adequately sensitive method with respect to *post-mortem* estimates of plaque burden, studies have shown significant variability across readers.^{29, 30} Findings from several studies indicate that use of quantitation could prove a helpful adjunct to visual interpretation,³¹⁻³⁴ similar to other areas of nuclear medicine involving PET imaging.^{35, 36} Quantitation of A β -PET images using SUVR derived from commercial software packages, for instance, has been shown to improve the accuracy of visual reads in clinically relevant cases.³⁷ This was also the case in the present study, where the addition of the A β -index improved prediction of histopathology based A β -status. The addition of an objective measure, like A β -index or SUVR, will probably be even more important in clinical practice, where many readers are not as experienced as those evaluating clinical PET images in academic research studies or clinical trials. The the A β -index—along with a cut-off indicating whether or not the scan is positive—could easily be incorporated into currently available commercial software.³⁸ Quantitation has also been shown to result in more consistent detection of early A β plaque pathology in CU older adults.³⁹ This finding in particular is of importance given that CU older individuals who are accumulating A β in the A β negative range—where visual

read alone is likely to prove insensitive—may prove a key target population for anti-A β clinical trials.⁴⁰

A fundamental step prerequisite to quantitation in PET is the spatial transformation of data into a common space (i.e. spatial normalization). SUVR values derived using the proposed principal component method were tightly correlated to those based on the dual-scan (MRI, PET) MRI driven approach used in SPM, indicating that accurate spatial normalization was achieved. Further, this method removes the need for a separate MRI scan; this is a highly desirable quality given that MR imaging is not always available as part of routine clinical workup and can be complicated by high rates of nonparticipation due to difficulty lying still during the examination, claustrophobia, contraindications such as pacemaker, metallic implants and other reasons.⁵ In addition, removing the need for MRI would decrease the burden placed on patients and caregivers, and a short computer tomography scan is often adequate to exclude secondary causes of cognitive impairment such as subdural hematoma and tumours and can be done in conjunction with the PET scan. In terms of clinical translation, additional studies are required to address whether the improvements in inter-reader agreement seen when adding SUVR to visual read of A β -PET images^{3,37} are also observed when using A β -index.

In both BioFINDER and ADNI, a range of A β -index values were observed for a given SUVR level. Despite identical SUVR levels, interpreted as indicating no difference in overall brain A β load, differences in the topography of A β pathology can be seen between subjects. These intersubject differences may explain the variability seen in A β -index for a given SUVR value. Though A β -index as a global metric of A β pathology may be of greatest interest from a clinical standpoint, further work addressing whether A β -index can in fact also provide information about A β pathology within different brain regions may be of interest, particularly with respect to the validation of PET based A β staging schemes^{26,41} and with an eye to testing

this approach with tau PET. The finding that A β -index values were only significantly different when comparing between advanced A β phases (3 vs 4 and 4 vs 5)—as for SUVR—however, is in line with earlier work showing that [^{18}F]flutemetamol PET primarily detects A β pathology in cases with advanced plaque pathology (i.e. A β phase \geq 4).⁴²

Two previous studies have used adaptive⁴³ and principal component derived templates⁴⁴ with A β -PET. In the study by Lundqvist et al.⁴³ using [^{18}F]flutemetamol PET, intercept and slope images were generated using linear regression; the slope image in combination with a weight is then used to generate a template. While the slope image is similar to our second principal component image, the principal component derived template appeared to provide greater accuracy.⁷ In the study by Fripp et al.⁴⁴ an adaptive template was generated using [^{11}C]PIB and spline based transformations for the normalization step; due to the use of splines, however, the computational time of their approach exceeds six hours, in contrast to an average processing time of \sim 20 seconds per subject using our method. Furthermore, a novel measure termed A β load (A β_{L}) was recently presented.²⁴ A metric of global A β burden, A β_{L} is calculated as a linear combination of two canonical images (nonspecific binding and a carrying capacity image representing the maximum possible concentration of A β).⁴⁵ Though these images and A β_{L} are conceptually similar, respectively, to our principal component images and A β -index, in contrast to the method by Gunn et al. our method does not require the use of MRI and is several orders of magnitude faster from a per subject computational standpoint.

Due to variability in the acquisition windows used for scanning, analysis methods and ROI selection, quantitatively expressed A β -PET outcome data cannot currently be directly compared. In an attempt to address this, a method was proposed whereby A β -PET values are standardized to a 100-point scale using a linear scaling procedure.^e The units of this scale are termed “Centiloids”, with 0 representing the average uptake in A β -negative subjects and 100,

the average in patients with mild-to-moderate AD.¹⁸ While the A β -index could be converted to Centiloids via the prescribed steps, the fact that it is independent of ROI definition suggests that it could be directly comparable between A β -PET tracers, without the need for conversion to the Centiloid scale. This would require, however, that a universal adaptive template be established by applying a principal component-based analysis to a data set comprising existing commercial A β -PET tracers. Future work is required to explore this possibility.

Strengths of the present study include the use of two different A β -PET tracers across three independent cohorts, a large sample size and the use of multiple standards of truth for defining A β -status. Certain limitations apply as well, however. First, cases of AD dementia or non-AD neurodegenerative disorders were not included. The close association of A β -index to SUVR across a range of values, however, indicates that the relationship between these metrics is governed by brain A β levels and is therefore independent of clinical diagnosis. As such, omitting these diagnostic groups is unlikely to have affected our results. Second, the patients in the Swedish BioFINDER study have been recruited in a consecutive fashion at three different memory clinics, with approximately 90% of these referred by primary care physicians. In the ADNI study, the patients were recruited from many different clinics, and thereby represent a more selected sample. Still, the results obtained for the A β -PET pathology accumulation index in both these studies are very similar. In light of findings showing that clinic-based cohorts—such as ADNI and, to a lesser extent, BioFINDER—might have a lower prevalence of infarcts and mixed pathologies,⁴⁶ further studies are required to validate the use of A β -index in community-based cohorts. Third, we did not examine the effect of atrophy on A β and SUVR. As neurodegenerative disorders such as AD are accompanied by progressive cortical atrophy, susceptibility to partial volume effects increases which, in the case of A β -PET, can diminish estimates of tracer retention.⁵⁴ While partial volume effects

were likely present to some degree in the BioFINDER and ADNI cohorts, the strong correlation between A β -index and SUVR across subjects suggests that these metrics were not differentially affected. We cannot exclude, however, that the strength of the association between A β -index and SUVR may be affected by atrophy in individual cases. Though fully quantitative measures (i.e. binding potential [BP_{ND}]) or distribution volume ratio [DVR]) would have been preferable over the use of SUVR—due the sensitivity of SUVR to cerebral blood flow induced changes in tracer kinetics⁴⁷—BP_{ND} and DVR require the use of dynamic data, which was not available. Lastly, in addition to the lack of *antemortem* diagnosis (beyond the presence or absence of dementia), there was considerable variation in the PET to *post-mortem* delay (scan-to-death time interval) in the end-of-life cohort; though this may have resulted in changes in A β burden not captured by the initial [¹⁸F]flutemetamol studies, our findings with the A β phases and prior work showing that the PET-to-death time interval did not affect the diagnostic performance of [¹⁸F]flutemetamol,⁴⁸ argues against this being the case.

Conclusions

Though the proposed A β -index showed a tight association to SUVR values and similar discriminative and predictive performance, it carries an advantage over SUVR in that it does not require the definition of target and reference regions. The A β -index may therefore prove simpler to implement in clinical settings. Further work is needed to address whether the A β -index could be implemented as a common measure across different A β tracers and analytical approaches, without the need for standardization to the Centiloid scale. An A β -index driven approach would require the availability of a hybrid template derived from all three commercially available A β tracers. This template is under development and will be the focus of future work.

Figure legends

Figure 1. Flow diagram providing an overview of the MRI-free normalization method

Shown are the steps required to generate the adaptive template and to spatially normalize the input A β -PET image. In the upper right corner, the role of the A β -index is illustrated: the first principal component image is combined with a weighted version of the second component image, yielding a template image optimal for the input image.

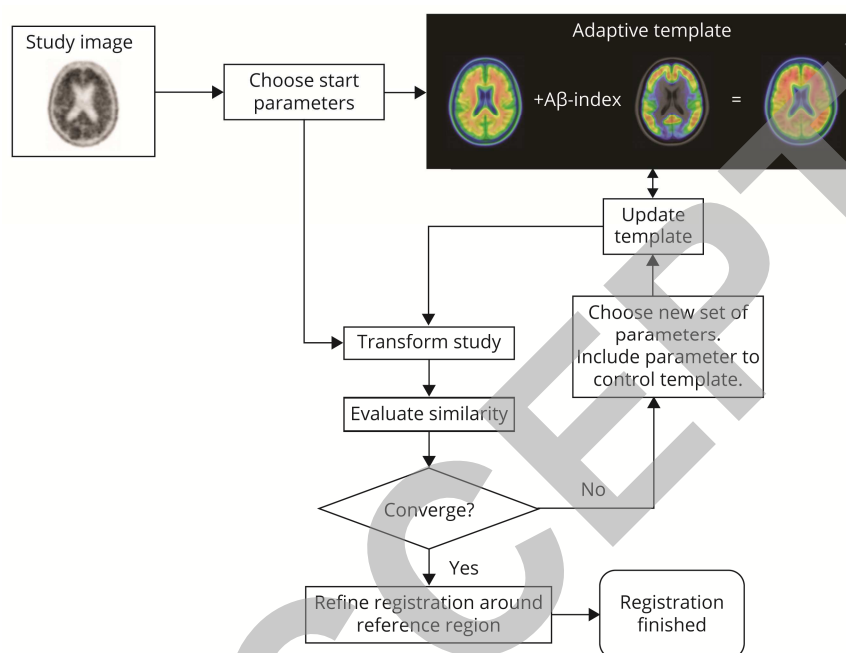


Figure 2. Scatterplots showing the relationship between A β -index and SUVR

The association between A β -index and SUVR is shown for [18 F]flutemetamol (BioFINDER cohort; A and C) and [18 F]florbetapir (ADNI cohort; B, D) PET, with A β status defined using visual read (A, B) and CSF A β 42/A β 40 (C, D) as standards of truth for A β status.

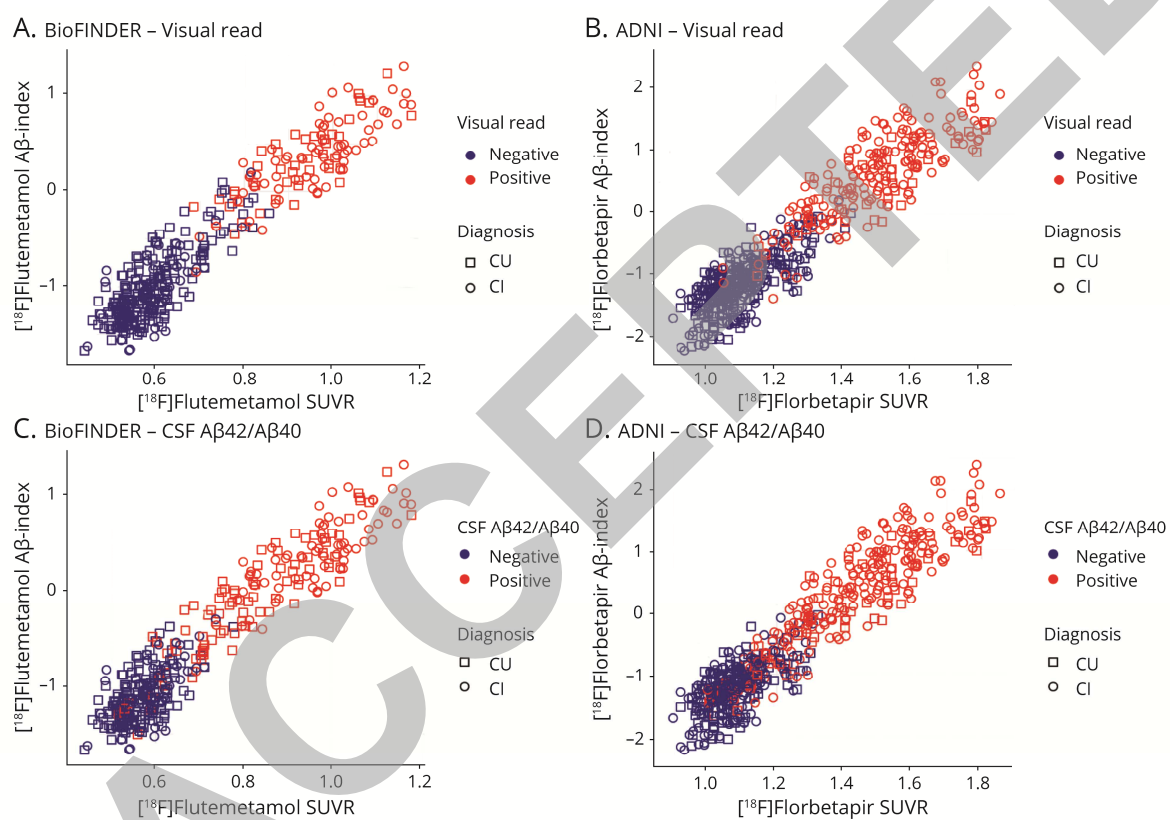


Figure 3. Receiver operating characteristic plots using A β -index and SUVR

Receiver operating characteristic plots for [18 F]flutemetamol (BioFINDER cohort; A, C) and [18 F]florbetapir (ADNI cohort; B, D) for distinguishing A β -negative and A β -positive subjects using visual read (A, B) and CSF A β 42/A β 40 (C, D) as standards of truth for A β status.

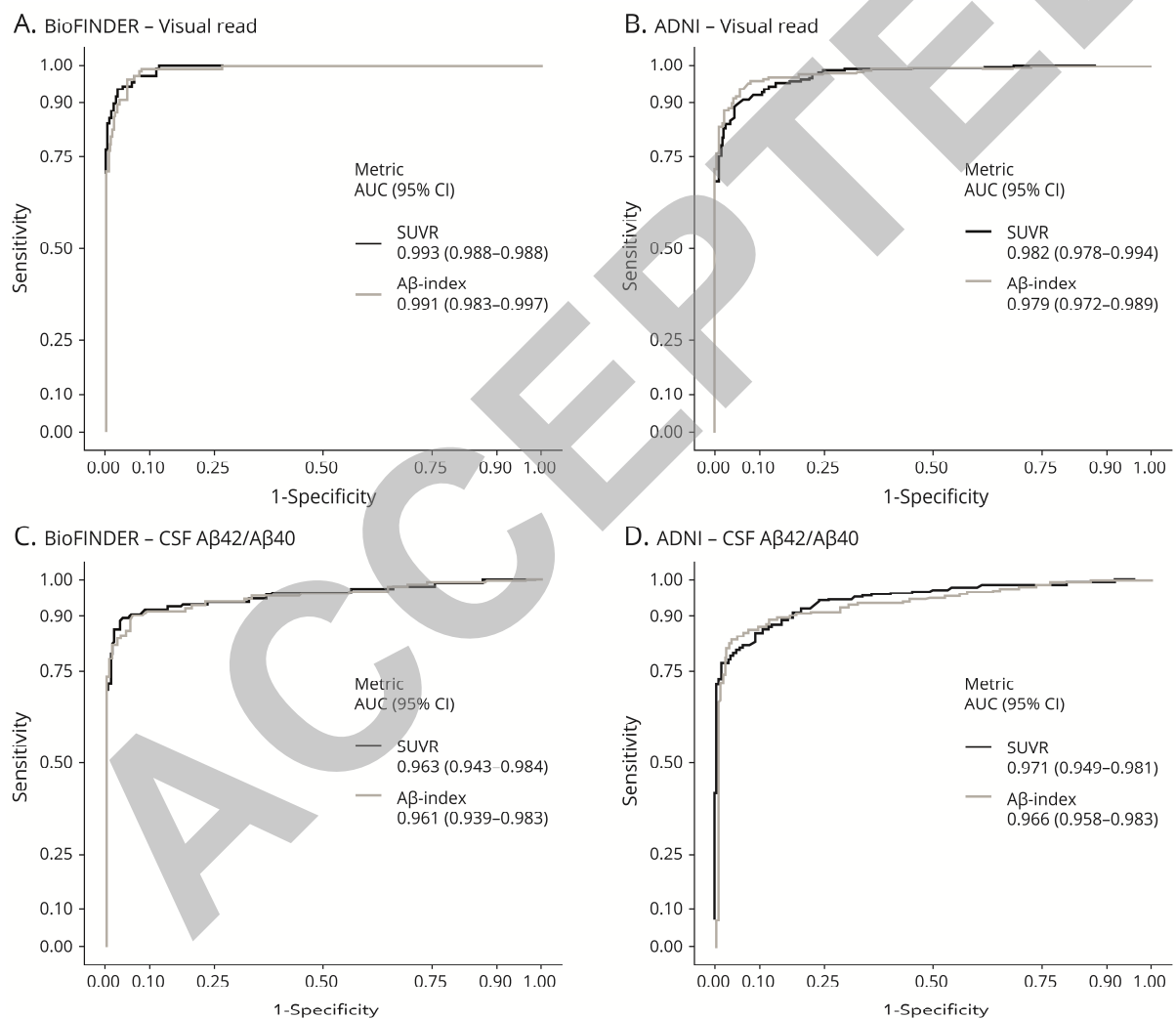


Figure 4. Findings from the [^{18}F]flutemetamol phase 3 end-of-life cohort

Receiver operating characteristic plots for distinguishing A β -negative and A β -positive subjects using the Bielschowsky silver stain score and visual read are shown in A and B, respectively.

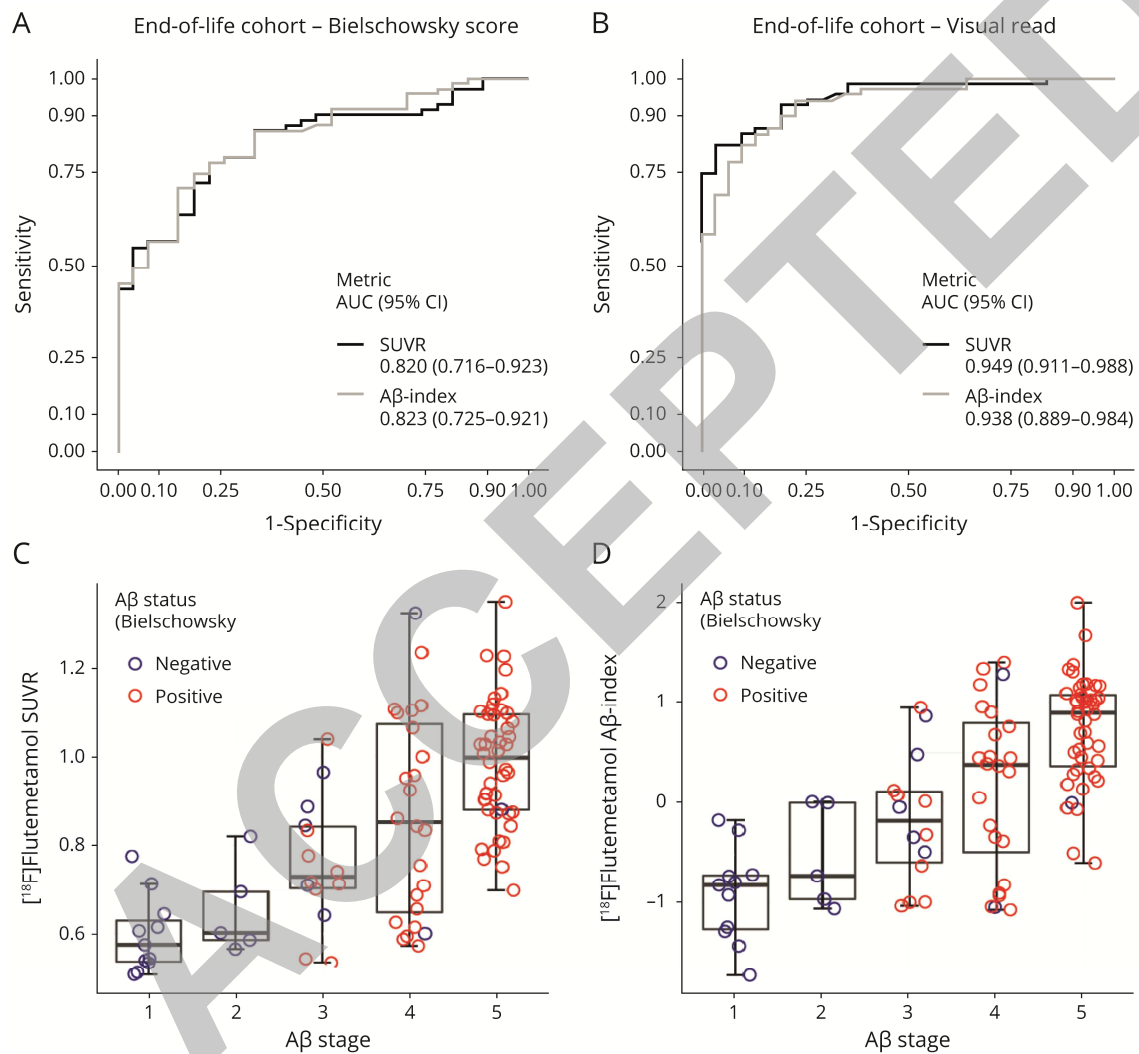


Table 1. BioFINDER and ADNI cohort characteristics

Characteristic	BioFINDER		ADNI	
	CU	CI	CU	CI
N	251	141	246	383
Age, y, mean (SD) [range]	71.92 (5.15) [59, 85]	70.91 (5.58) [60, 80]	72.16 (5.79) [56, 89]	71.05 (7.53) [55, 89]
Sex, Male/Female (% Male)	109/142 (43%)	88/53 (62%)	121/125 (49%)	208/175 (54%)
Education, years	12.17 (3.39)	11.14 (3.29)	16.82 (2.49)	16.3 (2.59)
MMSE score	28.79 (1.23)	27.26 (1.71)	29.05 (1.18)	28.11 (1.73)
APOE ϵ 4 + n %	91 (36%)	66 (47%)	51 (28%)	221 (58%)
CSF A β 42/A β 40 + n %	71 (28%)	88 (62%)	71 (29%)	211 (55%)
A β PET, visual read + n %	48 (19%)	79 (56%)	51 (21%)	185 (48%)
A β PET, SUVR	0.65 (0.15)	0.81 (0.21)	1.17 (0.19)	1.28 (0.24)
A β PET, A β -index	-0.79 (0.63)	-0.21 (0.78)	-0.82 (0.84)	-0.23 (0.95)

Table 2. End-of-life cohort characteristics

Characteristic	Dementia	No-dementia	All subjects
N	85	15	100
Age, y, mean (SD) [range]	82.71 (7.91) [60, 96]	78.40 (11.09) [60, 93]	82.10 (8.54) [60, 96]
Sex, Male/Female (% Male)	32/53 (38%)	5/10 (33%)	42/58 (58%)
Dementia, n %	85 (100%)	0 (0%)	85 (85%)
Bielschowsky silver stain	2.00 (0.79)	1.31 (1.13)	1.90 (0.88)
Thal A β phase, n, 1/2/3/4/5	6/4/10/21/44	5/1/4/2/3	11/5/14/24/46
Braak tau stage, n, I-II/III-IV/V-VI	14/17/50	4/8/1	18/25/51
CERAD, N/S/M/F	4/18/26/37	5/1/6/2	9/20/32/39
A β PET, visual read + n %	67 (79%)	5/10 (33%)	72 (72%)

A β PET, SUVR	0.88 (0.20)	0.79 (0.22)	0.87 (0.21)
A β PET, A β -index	0.31 (0.81)	-0.19 (0.89)	0.23 (0.84)
Scan-to-death, days	234.95 (215.48)	234.13 (189.26)	234.83 (210.86)

Abbreviations: N/S/M/F=None/Sparse/Moderate/Frequent

ACCEPTED

Appendix 1: Authors

Name	Location	Contribution
Antoine Leuzy, PhD	Lund University, Malmö, Sweden	Design and conceptualized study; analyzed the data; performed the statistical analyses. Drafted the manuscript for intellectual content
Johan Lilja, PhD	Lund University, Malmö, Sweden	Design and conceptualized study; analyzed the data; drafted the manuscript for intellectual content
Christopher J. Buckley, PhD	GE Healthcare Life Sciences, Amersham, UK	Design and conceptualized study; analyzed the data; drafted the manuscript for intellectual content
Rik Ossenkoppele, PhD	Lund University, Malmö, Sweden; VU University Medical Center, Amsterdam, The Netherlands	Major role in the acquisition of data; interpreted the data; revised the manuscript for intellectual content
Sebastian Palmqvist, MD, PhD	Lund University, Malmö, Sweden; Skåne University Hospital, Lund, Sweden	Major role in the acquisition of data; interpreted the data; revised the manuscript for intellectual content
Mark Battle, MSc	GE Healthcare Life Sciences, Amersham, UK	Major role in the acquisition of data; interpreted the data; revised the manuscript for intellectual content
Gill Farrar, PhD	GE Healthcare Life Sciences, Amersham, UK	Major role in the acquisition of data; interpreted the data; revised the manuscript for intellectual content
Dietmar R. Thal, MD	Leuven Brain Institute, Leuven, Belgium; UZ-Leuven, Leuven, Belgium	Design and conceptualized study; analyzed the data; drafted the manuscript for intellectual content
Shorena Janelidze, PhD	Lund University, Malmö, Sweden	Major role in the acquisition of data; interpreted the data; revised the manuscript for intellectual content

Erik Stomrud, MD, PhD	Lund University, Malmö, Sweden; Skåne University Hospital, Lund, Sweden	Major role in the acquisition of data; interpreted the data; revised the manuscript for intellectual content
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Oskar Hansson, MD, PhD	Lund University, Malmö, Sweden; Skåne University Hospital, Lund, Sweden	Design and conceptualized study; analyzed the data; drafted the manuscript for intellectual content

References

1. Seibyl J, Catafau AM, Barthel H, et al. Impact of Training Method on the Robustness of the Visual Assessment of 18F-Florbetaben PET Scans: Results from a Phase-3 Study. *J Nucl Med* 2016;57:900-906.
2. Buckley CJ, Sherwin PF, Smith AP, Wolber J, Weick SM, Brooks DJ. Validation of an electronic image reader training programme for interpretation of [18F]flutemetamol beta-amyloid PET brain images. *Nucl Med Commun* 2017;38:234-241.
3. Nayate AP, Dubroff JG, Schmitt JE, et al. Use of Standardized Uptake Value Ratios Decreases Interreader Variability of [18F] Florbetapir PET Brain Scan Interpretation. *AJNR Am J Neuroradiol* 2015;36:1237-1244.
4. Schmidt ME, Chiao P, Klein G, et al. The influence of biological and technical factors on quantitative analysis of amyloid PET: Points to consider and recommendations for controlling variability in longitudinal data. *Alzheimers Dement* 2015;11:1050-1068.
5. Tell GS, Lefkowitz DS, Diehr P, Elster AD. Relationship between balance and abnormalities in cerebral magnetic resonance imaging in older adults. *Arch Neurol* 1998;55:73-79.
6. Desikan RS, Segonne F, Fischl B, et al. An automated labeling system for subdividing the human cerebral cortex on MRI scans into gyral based regions of interest. *Neuroimage* 2006;31:968-980.
7. Lilja J, Leuzy A, Chiotis K, Savitcheva I, Sorensen J, Nordberg A. Spatial normalization of [(18)F]flutemetamol PET images utilizing an adaptive principal components template. *J Nucl Med* 2018.
8. Beach TG, Thal DR, Zhanette M, Smith A, Buckley C. Detection of Striatal Amyloid Plaques with [18F]flutemetamol: Validation with Postmortem Histopathology. *J Alzheimers Dis* 2016;52:863-873.
9. Thal DR, Beach TG, Zhanette M, et al. Estimation of amyloid distribution by [(18)F]flutemetamol PET predicts the neuropathological phase of amyloid beta-protein deposition. *Acta Neuropathol* 2018;136:557-567.
10. Ikonovic MD, Buckley CJ, Heurling K, et al. Post-mortem histopathology underlying beta-amyloid PET imaging following flutemetamol F 18 injection. *Acta Neuropathol Commun* 2016;4:130.
11. Palmqvist S, Zetterberg H, Blennow K, et al. Accuracy of brain amyloid detection in clinical practice using cerebrospinal fluid beta-amyloid 42: a cross-validation study against amyloid positron emission tomography. *JAMA Neurol* 2014;71:1282-1289.
12. Petersen RC, Aisen PS, Beckett LA, et al. Alzheimer's Disease Neuroimaging Initiative (ADNI): clinical characterization. *Neurology* 2010;74:201-209.
13. Landau SM, Mintun MA, Joshi AD, et al. Amyloid deposition, hypometabolism, and longitudinal cognitive decline. *Ann Neurol* 2012;72:578-586.
14. Jack CR, Jr., Bernstein MA, Fox NC, et al. The Alzheimer's Disease Neuroimaging Initiative (ADNI): MRI methods. *J Magn Reson Imaging* 2008;27:685-691.
15. Vandenberghe R, Van Laere K, Ivanou A, et al. 18F-flutemetamol amyloid imaging in Alzheimer disease and mild cognitive impairment: a phase 2 trial. *Ann Neurol* 2010;68:319-329.
16. Clark CM, Schneider JA, Bedell BJ, et al. Use of florbetapir-PET for imaging beta-amyloid pathology. *JAMA* 2011;305:275-283.
17. Johnson KA, Sperling RA, Gidyczin CM, et al. Florbetapir (F18-AV-45) PET to assess amyloid burden in Alzheimer's disease dementia, mild cognitive impairment, and normal aging. *Alzheimers Dement* 2013;9:S72-83.

18. Klunk WE, Koeppe RA, Price JC, et al. The Centiloid Project: standardizing quantitative amyloid plaque estimation by PET. *Alzheimers Dement* 2015;11:1-15 e11-14.
19. Bittner T, Zetterberg H, Teunissen CE, et al. Technical performance of a novel, fully automated electrochemiluminescence immunoassay for the quantitation of beta-amyloid (1-42) in human cerebrospinal fluid. *Alzheimers Dement* 2016;12:517-526.
20. Hansson O, Seibyl J, Stomrud E, et al. CSF biomarkers of Alzheimer's disease concord with amyloid-beta PET and predict clinical progression: A study of fully automated immunoassays in BioFINDER and ADNI cohorts. *Alzheimers Dement* 2018;14:1470-1481.
21. Korecka M, Waligorska T, Figurski M, et al. Qualification of a surrogate matrix-based absolute quantification method for amyloid-beta(4)(2) in human cerebrospinal fluid using 2D UPLC-tandem mass spectrometry. *J Alzheimers Dis* 2014;41:441-451.
22. Curtis C, Gamez JE, Singh U, et al. Phase 3 trial of flutemetamol labeled with radioactive fluorine 18 imaging and neuritic plaque density. *JAMA Neurol* 2015;72:287-294.
23. Mirra SS, Heyman A, McKeel D, et al. The Consortium to Establish a Registry for Alzheimer's Disease (CERAD). Part II. Standardization of the neuropathologic assessment of Alzheimer's disease. *Neurology* 1991;41:479-486.
24. Vemuri P, Whitwell JL, Kantarci K, et al. Antemortem MRI based STructural Abnormality iNDex (STAND)-scores correlate with postmortem Braak neurofibrillary tangle stage. *Neuroimage* 2008;42:559-567.
25. Palmqvist S, Janelidze S, Stomrud E, et al. Performance of Fully Automated Plasma Assays as Screening Tests for Alzheimer Disease-Related beta-Amyloid Status. *JAMA Neurol* 2019.
26. Thal DR, Rub U, Orantes M, Braak H. Phases of A beta-deposition in the human brain and its relevance for the development of AD. *Neurology* 2002;58:1791-1800.
27. Burnham KP, Anderson DR. Multimodel Inference: Understanding AIC and BIC in Model Selection. *Sociological Methods & Research* 2004;33:261-304.
28. Rabinovici GD, Gatsonis C, Apgar C, et al. Association of Amyloid Positron Emission Tomography With Subsequent Change in Clinical Management Among Medicare Beneficiaries With Mild Cognitive Impairment or Dementia. *JAMA* 2019;321:1286-1294.
29. Joshi AD, Pontecorvo MJ, Clark CM, et al. Performance characteristics of amyloid PET with florbetapir F 18 in patients with alzheimer's disease and cognitively normal subjects. *J Nucl Med* 2012;53:378-384.
30. Clark CM, Pontecorvo MJ, Beach TG, et al. Cerebral PET with florbetapir compared with neuropathology at autopsy for detection of neuritic amyloid-beta plaques: a prospective cohort study. *Lancet Neurol* 2012;11:669-678.
31. Camus V, Payoux P, Barre L, et al. Using PET with 18F-AV-45 (florbetapir) to quantify brain amyloid load in a clinical environment. *Eur J Nucl Med Mol Imaging* 2012;39:621-631.
32. Guerra UP, Nobili FM, Padovani A, et al. Recommendations from the Italian Interdisciplinary Working Group (AIMN, AIP, SINDEM) for the utilization of amyloid imaging in clinical practice. *Neurol Sci* 2015;36:1075-1081.
33. Kobylecki C, Langheinrich T, Hinz R, et al. 18F-florbetapir PET in patients with frontotemporal dementia and Alzheimer disease. *J Nucl Med* 2015;56:386-391.
34. Perani D, Schillaci O, Padovani A, et al. A survey of FDG- and amyloid-PET imaging in dementia and GRADE analysis. *Biomed Res Int* 2014;2014:785039.
35. Foster NL, Heidebrink JL, Clark CM, et al. FDG-PET improves accuracy in distinguishing frontotemporal dementia and Alzheimer's disease. *Brain* 2007;130:2616-2635.
36. Lin C, Itti E, Haioun C, et al. Early 18F-FDG PET for prediction of prognosis in patients with diffuse large B-cell lymphoma: SUV-based assessment versus visual analysis. *J Nucl Med* 2007;48:1626-1632.

37. Pontecorvo MJ, Arora AK, Devine M, et al. Quantitation of PET signal as an adjunct to visual interpretation of florbetapir imaging. *Eur J Nucl Med Mol Imaging* 2017;44:825-837.
38. Lilja J, Leuzy A, Chiotis K, Savitcheva I, Sorensen J, Nordberg A. Spatial Normalization of (18)F-Flutemetamol PET Images Using an Adaptive Principal-Component Template. *J Nucl Med* 2019;60:285-291.
39. Harn NR, Hunt SL, Hill J, Vidoni E, Perry M, Burns JM. Augmenting Amyloid PET Interpretations With Quantitative Information Improves Consistency of Early Amyloid Detection. *Clin Nucl Med* 2017;42:577-581.
40. McMillan CT, Chetelat G. Amyloid "accumulators": The next generation of candidates for amyloid-targeted clinical trials? *Neurology* 2018;90:759-760.
41. Mattsson N, Palmqvist S, Stomrud E, Vogel J, Hansson O. Staging beta-Amyloid Pathology With Amyloid Positron Emission Tomography. *JAMA Neurol* 2019.
42. Thal DR, Beach TG, Zhanette M, et al. [(18)F]flutemetamol amyloid positron emission tomography in preclinical and symptomatic Alzheimer's disease: specific detection of advanced phases of amyloid-beta pathology. *Alzheimers Dement* 2015;11:975-985.
43. Lundqvist R, Lilja J, Thomas BA, et al. Implementation and validation of an adaptive template registration method for 18F-flutemetamol imaging data. *J Nucl Med* 2013;54:1472-1478.
44. Fripp J, Bourgeat P, Raniga P, et al. MR-less high dimensional spatial normalization of 11C PiB PET images on a population of elderly, mild cognitive impaired and Alzheimer disease patients. *Med Image Comput Comput Assist Interv* 2008;11:442-449.
45. Whittington A, Sharp DJ, Gunn RN, Alzheimer's Disease Neuroimaging I. Spatiotemporal Distribution of beta-Amyloid in Alzheimer Disease Is the Result of Heterogeneous Regional Carrying Capacities. *J Nucl Med* 2018;59:822-827.
46. Schneider JA, Aggarwal NT, Barnes L, Boyle P, Bennett DA. The neuropathology of older persons with and without dementia from community versus clinic cohorts. *J Alzheimers Dis* 2009;18:691-701.
47. Carson RE, Channing MA, Blasberg RG, et al. Comparison of bolus and infusion methods for receptor quantitation: application to [18F]cyclofoxy and positron emission tomography. *J Cereb Blood Flow Metab* 1993;13:24-42.
48. Salloway S, Gamez JE, Singh U, et al. Performance of [(18)F]flutemetamol amyloid imaging against the neuritic plaque component of CERAD and the current (2012) NIA-AA recommendations for the neuropathologic diagnosis of Alzheimer's disease. *Alzheimers Dement (Amst)* 2017;9:25-34.

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