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Rates of β -amyloid deposition indicate widespread simultaneous accumulation throughout the brain

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Abstract

Amyloid plaque aggregation is a pathologic hallmark of Alzheimer's disease (AD) that occurs early in the disease. However, little is known about its progression throughout the brain. Using Pittsburgh Compound B (PIB)-PET imaging, we investigated the progression of regional amyloid accumulation in cognitively normal older adults. We found that all examined regions reached their peak accumulation rates 24–28 years after an estimated initiation corresponding to the mean baseline PIB-PET signal in amyloid-negative older adults. We also investigated the effect of increased genetic risk conferred by the apolipoprotein-E ϵ 4 allele on rates of amyloid accumulation, as well as the relationship between regional amyloid accumulation and regional tau pathology, another hallmark of AD, measured with Flortaucipir-PET. Carriers of the ϵ 4 allele had faster amyloid accumulation in all brain regions. Furthermore, in all regions excluding the temporal lobe, faster amyloid accumulation was associated with greater tau burden. These results indicate that amyloid accumulates near-simultaneously throughout the brain and is associated with higher AD pathology, and that genetic risk of AD is associated with faster amyloid accumulation.

Keywords

Aging; Amyloid; Tau; APOE; Alzheimer's Disease; Longitudinal

1. Introduction

Amyloid-beta ($A\beta$) plaques are a hallmark of Alzheimer's disease and are also present in cognitively normal older adults. By their 70s, almost a third of older adults without cognitive impairment have extensive amyloid accumulation (Jack et al., 2008). Cognitively unimpaired individuals with abnormal levels of $A\beta$ may have subtle brain and cognitive changes consistent with the presence of a “preclinical” phase of Alzheimer's disease (AD) (Sperling et al., 2011), and may be on an “Alzheimer's continuum” or have Alzheimer's disease based on biomarker categorization (Jack et al., 2018). Thus, identifying the earliest stages of $A\beta$ accumulation may be key to detecting who is at the highest risk for developing

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AD so that they can be followed closely, identified for clinical trials, and, if treatments become available, receive them in a timely manner.

$A\beta$ can be measured in vivo using cerebrospinal fluid (CSF) and positron emission tomography (PET) with $A\beta$ -sensitive radioligands. Often, studies assign participants to a “normal” or “abnormal” $A\beta$ group based on cut points using CSF or a global PET measure which encompasses frontal, temporal, parietal, and cingulate cortices. This may be missing early stages of $A\beta$ deposition when plaques may be present in only a subset of regions. Some studies have investigated the temporal progression of $A\beta$ pathology, but the results have been inconsistent. Neuropathological studies typically show that association cortex deposition precedes medial temporal lobe (MTL), primary sensorimotor cortex, and subcortical areas (Braak and Braak, 1991; Thal et al., 2002), but the cortical regions shown to incur the earliest $A\beta$ pathologic burden with neuroimaging vary in both cross-sectional (Cho et al., 2016; Grothe et al., 2017; Villeneuve et al., 2015; Yotter et al., 2013) and existing longitudinal studies (Guo et al., 2017; Jelistratova et al., 2020; Mattsson et al., 2019; Palmqvist et al., 2017).

The goal of this study was to use cross-sectional and longitudinal $A\beta$ -PET data to examine the regional changes in amyloid deposition when individuals are still cognitively normal. First, we replicated a method used previously (Cho et al., 2016) to show regions where elevated amyloid was seen cross-sectionally. For these cross-sectional analyses, we predicted that frontal, parietal, and lateral temporal regions would show evidence of earlier $A\beta$ accumulation than occipital, MTL, and sensorimotor regions. Identifying the earliest regions to display elevated $A\beta$ might allow us to detect cognitively normal individuals who are at high risk for AD when measures of whole-brain $A\beta$ burden are still below the threshold for positivity. This approach, combined with longitudinal measurements, could also reveal potential mechanisms underlying the spread of $A\beta$. For example, if amyloid deposits appear in 1 region years before depositing in functionally-connected regions, this may indicate spread occurs through functional networks. On the other hand, if $A\beta$ appears in multiple regions at once, these regions may share certain tissue properties that affect their vulnerability to $A\beta$ pathology (Whittington et al., 2018). After regional amyloid accumulation is staged, the relationship between regional amyloid accumulation and risk factors for AD and other pathologic markers of AD can be investigated.

One way to study longitudinal amyloid change is to examine the relationship between baseline $A\beta$ burden and rate of change in $A\beta$ burden, which is a quadratic function when measured globally (Blautzik et al., 2017; Jack et al., 2013; Jagust and Landau, 2021; Villemagne et al., 2013). Using the quadratic equation representing this association, we can calculate a sigmoidal relationship between $A\beta$ accumulation and time. The present study used this approach to investigate associations between regional slope and baseline global PET values to find the time course over which different regions reach their peak accumulation rates. The results from these longitudinal analyses were compared to a previous cross-sectional staging method (Cho et al., 2016). In addition, 2 previous studies have created a sigmoidal relationship between regional amyloid accumulation and time, one with modeled cross-sectional data (Whittington et al., 2018) and one with longitudinal data (Insel et al., 2020) and both show results that appear to be inconsistent with long-term

spread of $A\beta$ from one region to another. We, however, predicted that frontal, parietal, and lateral temporal regions would reach their peak accumulation rate before occipital and MTL regions, but that most association cortex regions would show similar timing in $A\beta$ accumulation.

In the process of examining rates of regional $A\beta$ accumulation, we also investigated the most important genetic factor involved in late onset AD, the Apolipoprotein E (*APOE*) genotype. Individuals with the $\epsilon 4$ allele are more likely to develop sporadic AD. In cognitively normal cohorts, $\epsilon 4$ carriers have been shown to have higher levels of $A\beta$ pathology than noncarriers (Fouquet et al., 2014; Resnick et al., 2015). However, studies of longitudinal $A\beta$ accumulation rates in $\epsilon 4$ carriers and noncarriers are inconsistent, with some finding no significant difference in accumulation rates and others showing higher accumulation rates in $\epsilon 4$ carriers than noncarriers. (Burnham et al., 2020; Lim and Mormino, 2017; Mishra et al., 2018; Resnick et al., 2015). It is unclear if $\epsilon 4$ carriers simply develop $A\beta$ earlier than noncarriers, but accumulate at the same rate, or if accumulation rate itself is faster in carriers versus noncarriers. Here we investigate the relationship between regional accumulation rates of $A\beta$ in $\epsilon 4$ carriers and noncarriers, controlling for baseline levels of pathology in each region. We predicted that $\epsilon 4$ carriers would have faster accumulation rates in regions found to accumulate $A\beta$ earlier (frontal, parietal, and lateral temporal regions) versus $\epsilon 4$ noncarriers.

Finally, we were interested in the association of $A\beta$ accumulation with another pathologic hallmark of AD, tau pathology. In contrast with $A\beta$, tau pathology accumulates following a stereotyped pattern beginning in the MTL (where high tau is common in older age), progressing to the surrounding inferolateral temporal lobes, and then spreading throughout the remaining cortex (Schöll et al., 2016; Vogel et al., 2020). Tau and $A\beta$ burden are positively correlated in animal (He et al., 2018) and human studies (Lockhart et al., 2017; Vemuri et al., 2017) and individuals who have high levels of both are more likely to develop dementia (Pascoal et al., 2017).

The mechanisms by which high levels of tau and $A\beta$ interact to create or contribute to cognitive decline are still unclear, especially since the regions susceptible to early tau, and $A\beta$ deposition do not overlap. While relationships between cross-sectional PET measures of $A\beta$ and tau have been explored (Lockhart et al., 2017), there are no data examining how the regional rates of $A\beta$ deposition may lead to regional tau deposition. Thus, we sought to investigate the relationship between $A\beta$ accumulation rates and cross-sectional tau-PET proximate to the final $A\beta$ measurement, within, and between regions. By creating a map of early $A\beta$ -depositing regions, we were able to investigate the relationships between regional accumulation rates and increased tau, both local, and distant. We predicted that MTL tau would be associated with rates of $A\beta$ accumulation in all regions, but that tau pathology outside the MTL would only be associated with accumulation in early $A\beta$ regions.

Therefore, our study seeks to increase understanding of how the time course of $A\beta$ accumulation varies by region, whether these accumulation rates differ in those with genetic risk for sporadic AD, and how accumulation rates in regions with different time courses associate with tau pathology in AD.

2. Materials and methods

2.1. Participants

We recruited 195 participants from the Berkeley Aging Cohort Study (Mormino et al., 2009). All participants were cognitively normal, over 60 years of age, and had at least 1 [¹¹C] Pittsburgh compound B (PIB) Positron emission tomography (PET) scan for measurement of $A\beta$. A subset of 106 participants had more than 1 PIB-PET scan and were used for longitudinal analyses. Eighty-eight of these 106 participants had at least 1 [¹⁸F] Flortaucipir (FTP) PET scan to measure tau pathology proximal to their final PIB-PET scan. All but 10 participants had FTP and PIB-PET scans on the same day and the median date difference was 0 days. Across all scans, the median difference between the PIB-PET and MRI scan was 8 days. Participant characteristics can be found in Table 1. Amyloid positive individuals were selectively recruited for longitudinal studies, producing an uncharacteristically high rate of amyloid positivity.

2.2. Image acquisition

[¹¹C] PIB was synthesized at Lawrence Berkeley National Lab (LBNL) Biomedical Isotope Facility as previously described (Mathis et al., 2003). PIB-PET imaging was carried out on 1 of 2 scanners, the ECAT EXACT HR or the BIOGRAPH PET/CT Truepoint 6. Previous studies have shown that there is no significant difference in distribution volume ratios (DVRs) between the 2 scanners (Elman et al., 2014).

Details of PIB-PET scan acquisition have been previously outlined (Lockhart et al., 2017). Subjects received an intravenous injection of PIB (~15 mCi), at the beginning of 90 minutes of dynamic acquisition frames. The data were binned into 35 frames (4 × 15 seconds, 8 × 30 seconds, 9 × 60 seconds, 2 × 180 seconds, 10 × 300 seconds, and 2 × 600 seconds). A transmission or CT scan was acquired immediately before the emission scan for attenuation correction.

[¹⁸F]FTP-PET was synthesized at LBNL based on a protocol provided by Avid Radiopharmaceuticals as previously described (Schöll et al., 2016). FTP scans were done on the BIOGRAPH scanner. Data acquired from 80 to 100 minutes post-injection were binned into 4 × 5-minute frames during reconstruction and used in subsequent analyses.

All PET images were reconstructed using an ordered subset expectation maximization algorithm with weighted attenuation, scatter correction, and smoothed with a 4mm Gaussian kernel. Participants also underwent MRI on a 1.5 Tesla Siemens Magnetom Avanto MRI scanner at LBNL. A FreeSurfer (FS; <http://surfer.nmr.mgh.harvard.edu/>) version 5.3 segmentation of a T1-weighted magnetization prepared rapid gradient echo (MPRAGE) scan (TR/TE = 2110/3.58 ms, FA = 15°, 1 × 1 × 1 mm resolution) was used to parcellate PET scans into regions of interest (ROIs) for analysis of PET images.

2.3. Image processing

2.3.1. PIB-PET—PIB-PET data were realigned, resliced, and coregistered to correspond to their closest structural scan from the 1.5 T MRI. There was a median of 7 days between

the scans (range 0–767 days). Logan graphical analysis was used to calculate the DVR values as the slope of frames 35–90 minutes post-injection with a cerebellar reference region (Logan et al., 1996; Price et al., 2005). The structural MRI corresponding to each PET scan was processed with FS to derive ROIs in native space for each subject using the Desikan-Killiany atlas. The PET image was coregistered to the MRI, allowing us to calculate mean regional DVR values for each FS ROI. The global cortical PIB measure was calculated in native space for each participant (Mormino et al., 2011). In addition, the following FS regions were averaged to form larger composite cortical regions: lateral parietal (superior parietal, supramarginal, inferior parietal), medial parietal (precuneus, isthmus cingulate, posterior cingulate, paracentral), lateral frontal (rostral middle frontal, caudal middle frontal, pars orbitalis, pars triangularis, pars opercularis), medial frontal (superior frontal, caudal anterior cingulate, rostral anterior cingulate), orbitofrontal (OFC; frontal pole, medial and lateral orbitofrontal) sensorimotor (precentral and postcentral), inferior lateral temporal (fusiform, inferior and middle temporal), superior lateral temporal (superior temporal, transverse temporal, banks of the superior temporal sulcus), insula (insula), MTL (entorhinal and parahippocampal) and occipital (lateral occipital, lingual, cuneus, pericalcarine).

2.3.2. FTP-PET—FTP Standardized Uptake Value Ratio (SUVR) images were generated based on mean tracer uptake 80–100 minutes post-injection using an inferior cerebellar gray matter reference region (Baker et al., 2017a). Like the PIB-PET data, the images were coregistered, and resliced to each subject's 1.5 Tesla MR scan closest to the FTP scan. The scans occurred, on average, 30 days apart. SUVR images were partial volume corrected using the Geometric Transfer Matrix approach on FreeSurfer-derived values (Baker et al., 2017b; Rousset et al., 1998). The same composite ROIs were calculated for FTP as for PIB.

2.4. Statistical analysis

Statistical Analyses were conducted using MATLAB version 9.4 (R2018a; MathWorks, Inc, Natick, Massachusetts) and R version 3.4 (<https://www.R-project.org/>), including the packages *nlme* for analyses and *ggplot2* and *ggseg3d* for data visualization.

2.4.1. Cross-sectional PIB staging analyses—We used cross-sectional data to infer a pattern of temporal progression of $A\beta$ deposition, similar to methods used by others in staging $A\beta$ pathology (Cho et al., 2016) in order to determine the ordering of positivity between regions, and to allow for comparison with the longitudinal methods. Mean and standard deviation values for each region were calculated in the subset of participants below the cutoff for PIB positivity ($N = 122$; $DVR < 1.065$; Villeneuve et al., 2015). These values were used to Z-score each region's data for the entire sample. For each region, the number of participants with a Z-score > 2.5 was counted, and regions were ranked from those with the highest proportion of individuals with Z-scores > 2.5 to those with the lowest. Ties were broken by calculating the mean value of the Z-scores, with higher Z-score means leading to a higher ranking.

2.4.2. Longitudinal PIB staging analyses—A linear function of DVR versus time was used to characterize the annual rate of change for subsequent analyses that minimized

distance from the regression line (Leal et al., 2018). Next, linear and quadratic models were fit in R using annual accumulation rate for each region (along with global PIB slope) and baseline global PIB values. In the linear model (slope ~ BL global PIB), regional slope was the dependent variable, and global PIB was the independent variable. The quadratic model included the same parameters, but also included a baseline global PIB squared term (slope ~ BL global PIB + BL global PIB²). For regions where both baseline global PIB terms were significantly associated with slope, Aikake Information Criterion (AIC), and residual sum of squares were used to compare the fit.

For regions where the quadratic model was a better fit, the maximum of the quadratic equation (created using the estimates from the model) could be found algebraically: for the function $f(x) = ax^2 + bx + c$, the maximum point is $(-b/(2a), f(-b/(2a)))$. The x coordinate represents the global PIB DVR at which this region's PIB accumulation rate is fastest. Because all equations used baseline global PIB DVR as the independent variable, the global DVRs at which each region reaches its peak accumulation can be directly compared.

We next calculated the annual change in global DVR starting from the mean baseline global PIB DVR for PIB negative participants (DVR = 1.02). Annual change was calculated using the quadratic function through a time course of 50 years (Jagust and Landau, 2021). We also converted DVR values to Centiloids (CLs), a standardized scale allowing for comparison of results using different A β -PET radiotracers (Klunk et al., 2015). This approach enabled us to determine when accumulation in a region starts to decelerate.

2.4.3. PIB-APOE associations—Participants were categorized as carriers or noncarriers of the APOE $\epsilon 4$ allele. Because noncarrier longitudinal accumulation data did not fit quadratic models, linear mixed effects models (LME models) carried out using the R package *nlme* were used for these analyses to compare the linear trajectories of the carriers and noncarriers. In these models, PIB-PET was the dependent variable, with fixed effects APOE status, age, and sex, all interacting with time. The random effects were (1+time|subject) which modeled a subject-specific intercept and allowed the results to vary by subject over time.

2.4.4. Amyloid-tau associations—Relationships between PIB-PET accumulation rates and FTP-PET within and across ROIs were examined using LMEMs. In these models FTP-PET was the dependent variable, with fixed effects PIB-PET, age, and sex as well as their interactions with time. The random effects were (1+time|subject). Local and distant analyses were run for all regions, as well as analyses examining the association between global PIB accumulation rate, and regional tau. All statistics are reported at a liberal threshold of $p < 0.05$, uncorrected.

3. Results

3.1. Cross-sectional results

The results of the cross-sectional analysis can be found in Table 2. There were small differences in the proportions of participants with abnormal values in frontal and parietal

cortex. Regions with the lowest proportion of abnormal subjects were sensorimotor, MTL, and occipital.

Fig. 1 shows these results graphically for each participant. Global PIB-PET captured the greatest number of people with elevated $A\beta$ pathology. Furthermore, most participants were either above the cut points in all or nearly all regions or were not above the cut point in any region. Taken together with the finding that global PIB had the most participants above a Z-score of 2.5, this indicates that although some individuals show regionally specific PIB-PET signal, levels of PIB do not considerably differ by region.

3.2. Longitudinal results

Quadratic Model Fit—Quadratic models better fit the data in 8 of 11 regions and global PIB. The global PIB value at which each region reaches its fastest rate of accumulation, as well as the average rate of accumulation for each region, can be seen in Table 3. Neither the fastest accumulating (OFC; 0.014 DVR/year) or the slowest accumulating (MTL; 0.003 DVR/year) fit a quadratic. The first region to reach its peak accumulation rate was lateral parietal. Insula was the final one, with a difference in DVR of 0.13, CL value of 17, and an estimated time difference of 2–4 years. The quadratic fits for a subset of regions are plotted in Fig. 2.

3.2.1. Regional PIB time course—Given a starting point of the mean global PIB value for PIB negative participants (DVR = 1.02; CL = 4), it would require 8–9 years to reach the threshold for PIB positivity (DVR = 1.065; CL = 10). The time course of global PIB accumulation can be seen in Fig. 3, with the points at which select regions reach their peaks superimposed on the global PIB graph.

3.3. PIB-APOE relationships

In all models, APOE status was significantly associated with PIB-PET, both in its own term, and interacting with time. Visualization showed that $\epsilon 4$ carriers had higher PIB-PET at baseline, and it increased at a slightly faster rate. Results of these models are listed in Table 4 and a subset of associations between $\epsilon 4$ carrier status and regional slope are shown in Fig. 4.

3.4. Tau-amyloid relationships

Relationships between longitudinal PIB accumulation and FTP looked similar, regardless of the PIB region in LMEs (Fig. 5). Local relationships were only seen between PIB accumulation rates and FTP in the lateral parietal ($t [157] = 2.28, p = 0.024$), superior lateral temporal ($t [157] = 2.19, p = 0.029$) and occipital regions ($t [157] = 2.70, p = 0.0078$). For most PIB regions, accumulation rate was most strongly associated with temporal FTP, especially lateral temporal regions, and lateral parietal FTP. PIB accumulation rates measured globally, and in medial parietal and OFC, were most strongly associated with FTP in temporoparietal brain regions, and also were associated with a wider distribution of FTP. Notably, these PIB regions, along with global PIB, were the 4 PIB measures with the highest percent involvement of PIB in cross-sectional methods, and among the fastest accumulation rates.

Furthermore, PIB accumulation in the medial, and inferior lateral temporal lobe was not associated with FTP in any region. Both regions had lower percent involvement of PIB in cross-sectional methods; inferior lateral temporal was one of the last regions to reach its peak accumulation rate in the quadratic analyses.

4. Discussion

In this study, we investigated global rates of regional $A\beta$ accumulation, and examined whether dynamics of $A\beta$ accumulation in different brain regions differed from one another. This focus on both global and regional rates of $A\beta$ accumulation allowed us to determine another measure of regional vulnerability to $A\beta$ and permitted us to see how regional rates of $A\beta$ accumulation related to APOE genotype and tau deposition. As anticipated, cross-sectional results showed that global PIB-PET signal captured the most individuals with elevated $A\beta$. Those with high levels of PIB-PET in one region were likely to have high PIB-PET in most regions, but small differences between regions indicated that high levels of $A\beta$ were most common in frontal and parietal regions, and least common in temporal cortex (particularly the MTL), the occipital lobe, and sensorimotor cortex. In longitudinal analyses, all investigated regions reach their peak accumulation rate within a span of 4 years (24–28 years after accumulation began), indicating that spread from one region to another is unlikely, although, as discussed below, this inference is subject to some limitations. Compared to noncarriers, APOE $\epsilon 4$ carriers had higher baseline $A\beta$, and faster accumulation in all regions even after controlling for regional baseline $A\beta$. Lastly, faster $A\beta$ accumulation was associated with elevated tau in temporal and parietal regions. These relationships were stronger and more widespread in the brain regions that deposited $A\beta$ more rapidly, and no relationships were seen between $A\beta$ accumulation in MTL or inferior lateral temporal regions and tau pathology in any region. Together, these results further our understanding of the regional dynamics of $A\beta$ accumulation in unimpaired older adults and show how the rate of accumulation relates to genetics and tau pathology.

4.1. Regional amyloid deposition

In the present study, there were regions that were more likely to harbor amyloid deposition than others, but most regions seemed to accumulate amyloid at a similar rate. Cross-sectionally, frontal and parietal regions were more likely to harbor elevated $A\beta$, and MTL sensorimotor, and occipital regions were less likely. Elevated $A\beta$ in one region, however, was typically indicative of elevated $A\beta$ throughout much of the cortex. These results are consistent with autopsy studies, in which frontal, parietal, and lateral temporal regions consistently show $A\beta$ plaques before MTL and primary sensorimotor cortex (Braak and Braak, 1991; Thal et al., 2002). Cross-sectional $A\beta$ -PET data show that deposition of $A\beta$ in the sensorimotor cortex, occipital lobe, and medial temporal regions is less at early stages of AD (Cho et al., 2016; Grothe et al., 2017; Yotter et al., 2013). Our findings using the methods from Cho et al. (2016) largely replicate their results and add to a body of literature showing elevated $A\beta$ in neocortical association areas with fewer individuals harboring high $A\beta$ in MTL, sensorimotor, and occipital regions.

Cross-sectional results, however, do not address whether brain regions accumulate $A\beta$ at different rates. In our longitudinal analyses, we found that all regions reached their peak accumulation rates 24–28 years after accumulation begins. A benefit of the present study is that, since accumulation in all regions was fitted to a quadratic with baseline global $A\beta$, we could directly compare the timing of accumulation. We found that, although some regions might accumulate $A\beta$ slightly earlier or more quickly, the temporal differences between regional rates of $A\beta$ accumulation are small. Many studies of regional $A\beta$ have created staging schema using various methods. Some studies use cutoffs to identify regions that are elevated more often. One such study identified basal portions of the temporal lobe, the anterior cingulate, and parietal operculum as the earliest regions of deposition, with large portions of the frontal, parietal, and temporal lobes following in the second stage (Grothe et al., 2018). Although this staging method was created with cross-sectional data, it has been verified longitudinally (Jelistratova et al., 2020; Teipel et al., 2020). Others identify participants as $A\beta$ accumulators or non-accumulators with the help of CSF measurements of $A\beta$, and compare where $A\beta$ is highest, or where it is increasing fastest in accumulators versus non-accumulators. In these studies, earliest-stage $A\beta$ appeared in the medial parietal lobe, medial frontal, and orbitofrontal cortex, with one additionally identifying lateral temporal regions (Mattsson et al., 2019; Palmqvist et al., 2017). Studies have also investigated either regions where participants have a high $A\beta$ burden compared to their global $A\beta$ value, or simply regions where either baseline $A\beta$ burden is highest or where $A\beta$ burden is increasing the fastest. Highest baseline values and fastest rates of change are often seen in the precuneus, anterior and posterior cingulate, and lateral frontal regions (Guo et al., 2018, 2017; Insel et al., 2020). What is typically consistent, however, is that medial temporal, occipital, and sensorimotor regions are among the final regions to become involved (Grothe et al., 2017; Guo et al., 2018, 2017; Mattsson et al., 2019; Palmqvist et al., 2017). Previous studies have often used clinically diverse cohorts including cognitively impaired participants. The wide array of regions that may define early stages of $A\beta$ deposition may be due to demographic differences and/or PET radiotracer properties magnifying small differences in different cohorts.

This study is largely consistent with previous studies modeling the time course of $A\beta$ over a sigmoidal trajectory. Important time points found in this paper and 3 others can be found in Table 5. Besides the present study, all other papers included cognitively impaired patients in their analyses (Jack et al., 2013; Jagust and Landau, 2021; Villemagne et al., 2013), and Villemagne et al. used mean PIB rather than baseline PIB. Despite these differences, the temporal patterns are surprisingly similar. From a starting point of the mean $A\beta$ burden of CNs, in the present study it would take 28–29 years to reach a typical $A\beta$ burden for AD patients (equivalent to a CL value of 100), versus 28 years in ADNI (Jagust and Landau, 2021) and 31 years in AIBL (Villemagne et al., 2013). The time course in AIBL is very similar to the present study, with only the time to the $A\beta$ positivity threshold more than approximately 2 years longer (12 years vs. 8–9 years). In ADNI the starting point for the sigmoid is the same as the present study (4 CL) and the overall time course is very similar, but the amount of time it takes to reach the peak accumulation rate from the $A\beta$ positivity threshold is much shorter (3 years vs. 17–18 years) and the time to reach an average AD patient value from that peak is much longer (18 years vs. 3–4 years). This may be due

to the high number of cognitively impaired participants in the ADNI study (>60%) along with the requirement that only controls with positive amyloid-PET slopes were included. Data from the Mayo Clinic Study of Aging (MCSA) are somewhat difficult to compare because the starting SUVR for inclusion was higher. Perhaps for this reason, timing from a positive scan (an SUVR of 1.5) to a value typical of an AD patient required only about 14 years. Across all studies the most variable period of accumulation seems to be from the $A\beta$ positivity threshold to the peak, which included values of 3, 6.6, and 18 years. Differences may be related in part to $A\beta$ tracer used, or methodological differences in PET processing (such as the chosen reference region). This is a dynamic phase of $A\beta$ accumulation and the inverted-U relationship is variable across studies, and any time course is relative to a study-defined starting point, so differences likely also reflect cohort composition.

Two papers have modeled sigmoidal trajectories for regional $A\beta$ (Insel et al., 2020; Whittington et al., 2018). As in this report, which modeled trajectories of regional accumulation relative to global $A\beta$ so regional time courses can be compared on the same time scale, these results found limited differences in accumulation between regions. Insel and colleagues modeled accumulation of $A\beta$ in different brain regions as a function of estimated disease time, and found that some regions, such as the precuneus and posterior cingulate, begin with high uptake and continue to increase in $A\beta$ -PET uptake over time; other regions (lateral OFC, isthmus cingulate, inferior parietal) start high but increase more slowly; in this report, all regions appear to reach their inflection point around the same time. This is similar to the findings of an earlier paper modeling accumulation with cross-sectional data, which concluded that amyloid accumulation begins simultaneously throughout the brain and that all regions reach their peak accumulation rates around the same time, with differences in the rate of accumulation based on carrying capacity (Whittington et al., 2018). This may help explain the somewhat surprising finding that sensorimotor cortex was one of the regions to reach its peak accumulation rate earlier despite relatively low levels of $A\beta$. This contrast between rapid accumulation rates and low values of $A\beta$ may reflect 2 distinct processes, such that vulnerability may reflect either the rate of accumulation or the resulting amount of pathology.

The overall similarities of our results with these previous studies, despite using only cognitively normal older adults, is notable. The time course appears to be elongated in the present study, which is not surprising given the entirely cognitively normal sample. The trajectory may be longer than in a sample of individuals who either have probable AD dementia or are likely to develop it (those with MCI, cognitive deficits, high $A\beta$, accelerating $A\beta$ accumulation, etc.). However, it is useful to investigate the time course of $A\beta$ in a cognitively normal sample as well. Many of these individuals are amyloid positive and so are considered to be on the Alzheimer's continuum; while we did not dichotomize tau outcomes, some would also be classified as having biomarker evidence of Alzheimer's disease according to the current research framework (Jack et al., 2018). Understanding the earliest stages of $A\beta$, before cognitive impairment, may be crucial to early treatment of the disease, and this time course would be harder to parse if patients were included in the model.

Overall, these results indicate that significant inter-region spread of $A\beta$, through functional connections or otherwise, is not likely, given the short time differences between each

region's peak accumulation rate. This seems especially unlikely when taking into account the variability in early-stage regions seen *post mortem* (Braak and Braak, 1991; Thal et al., 2002) and PET studies of regional A β (Guo et al., 2018; Mattsson et al., 2019; Sakr et al., 2019). This is also consistent with the frequent finding that in large samples, A β positive and A β negative participants dominate, with few intermediate individuals, consistent with widespread and rapid increases in A β in vulnerable individuals and regions. This stands in contrast with tau pathology, which is mainly restricted to MTL regions in older adults without cognitive impairments, seeming to spread to nearby temporal lobe locations and functionally-connected regions in the presence of higher A β over longer time periods and through documented networks of connectivity (Adams et al., 2019; Franzmeier et al., 2020; Schöll et al., 2016; Vogel et al., 2020). However, we recognize that our models depend on A β that can be visualized with PET. It is possible that soluble forms of A β , or even fibrillar forms at very low and undetectable levels, may spread through connectivity many years before signal can be detected with a PET scan. Furthermore, it is possible that network-based spread could happen extremely quickly.

It seems most likely that regions which accumulate A β earlier share factors that make them vulnerable to aggregated A β peptide accumulation. They may, for example, have similar properties. In a study of regional gene expression, higher expression of the A β precursor protein (APP) gene, and reduced expression of genes involved in protein synthesis and mitochondrial respiration are associated with higher levels of A β -PET (Grothe et al., 2018). Patterns of regional metabolism are also implicated based on data showing that amyloid accumulating regions are involved in aerobic glycolysis (Vlassenko et al., 2010) and that they have lifelong elevated levels of glucose metabolism (Oh et al., 2016). These genetic and metabolic factors are likely to drive shared vulnerability resulting in A β aggregation appearing almost simultaneously in susceptible regions.

4.2. APOE E4-regional A β accumulation relationships

In this study, we also found that carriers of the APOE ϵ 4 allele had faster accumulation of A β than noncarriers in all brain regions, adjusting for the difference in baseline A β . Our results are consistent with studies that report faster rates of accumulation in carriers (Burnham et al., 2020; Lim and Mormino, 2017; Mishra et al., 2018), although there are discrepancies (Lopresti et al., 2020; Resnick et al., 2015) Burnham et al (2020). and Lim and Mormino (2017) both noted effects of APOE ϵ 4 genotype early in the stage of A β deposition, prior to reaching an amyloid positivity threshold. Thus, it is possible that the ϵ 4 allele shifts the onset of A β deposition by subtly increasing rates prior to reaching an amyloid positive stage.

4.3. Regional amyloid-tau relationships

Longitudinal associations between A β accumulation and tau mirrored the relationships found in our previous report with cross-sectional data (Lockhart et al., 2017) and were also similar to the pattern of elevated tau seen in studies investigating longitudinal global A β and tau (Tosun et al., 2017). Faster PIB accumulation in most regions was associated with cross-sectional FTP-PET in a relatively limited group of brain regions, with strongest relationships seen between PIB in multiple regions and FTP in the temporal lobe, where pathologic

tau is known to accumulate early in AD (Braak and Braak, 1991; Johnson et al., 2016; Schöll et al., 2016). Nevertheless, there were some regional associations between rates of $A\beta$ accumulation and tau deposition, with no association between $A\beta$ accumulation in inferior lateral and medial temporal lobes and tau pathology, and stronger associations between $A\beta$ accumulation in faster accumulating regions (OFC and medial parietal cortex) and tau pathology. The fact that regional $A\beta$ accumulation was associated with a similar pattern of tau—especially temporal tau—is consistent with interpretations of prior cross-sectional data indicating that tau and $A\beta$ pathology start in different regions, and that the location of $A\beta$ accumulation bears non-specific relationship to the location of tau pathology both within and outside the temporal lobe. The similar findings for both cross-sectional and longitudinal $A\beta$ accumulation raise crucial unanswered questions about how the widespread pattern of $A\beta$ accumulation drives a relatively more focal deposition of tau pathology.

4.4. Limitations

There are some limitations to the present study. The sample size is relatively small, compared to some studies. About half of the participants have only 2 PIB-PET time points, whereas it is optimal for longitudinal analyses to have 3 or more time points. In addition, all participants in the cohort are cognitively unimpaired. This is partially a strength and is helpful for characterizing early pathologic changes and as previously mentioned, verifying that $A\beta$ accumulation in CN older adults follows a quadratic shape. However, it has drawbacks, such as limiting the number of high- $A\beta$ $\epsilon 4$ noncarriers. Our ability to extrapolate to the full spectrum across the natural history of AD is thus limited.

Aggregating the smaller, FreeSurfer-defined regions into larger regions is another potential limitation to the study. The data for small regions were noisy and did not typically follow a sigmoidal trajectory. Perhaps in a larger or more cognitively diverse cohort, sigmoidal functions could be fit for smaller regions. Similarly, the ideal way to compare $\epsilon 4$ carriers and noncarriers would be to fit separate quadratics. The noncarriers in this sample, however, did not fit a quadratic probably because of lower levels of $A\beta$ in the noncarriers in particular. In a larger sample, or one with impaired individuals, there may be enough noncarriers with high accumulation rates, which would allow for the fitting of 2 separate quadratics.

In the $A\beta$ -tau analyses, we chose to use a liberal threshold of $p < 0.05$, not correcting for multiple comparisons, to visualize our results. This analysis was meant primarily to show the patterns of elevated tau regions in relation to regional accumulation of $A\beta$, rather than to directly compare the extent of tau deposition associated with accumulation of $A\beta$ in each region. Because of this liberal threshold, we cannot claim from these analyses that accumulation of $A\beta$ in one region leads to higher tau than any other region. In fact, though, the results again indicate to us greater regional similarities than differences in the relationships between where $A\beta$ increases over time, and where tau deposits.

Finally, it is well known that $A\beta$ -PET imaging is limited by the fact that it can only show us the location of $A\beta$ plaques, which follow an earlier, oligomeric form of $A\beta$ that is considered more toxic (Sakono and Zako, 2010; Salahuddin et al., 2016). It is possible that the reason for inconsistencies in “early regions” or findings such as those in the present study that amyloid appears to arise near-simultaneously in many brain regions, is due to

PETs inability to image the very early phases of $A\beta$ accumulation. However, there is currently no method to image oligomeric $A\beta$.

5. Conclusions

This study indicates that $A\beta$ accumulates nearly simultaneously throughout the brain in cognitively normal older adults. Global $A\beta$ accumulation follows a sigmoidal trajectory over time, and the time of peak accumulation of $A\beta$ of different brain regions can be placed on that trajectory within a 4-year span, peaking approximately 24–28 years after participants reach the mean baseline $A\beta$ burden of cognitively normal, $A\beta$ -negative older adults. Furthermore, faster rates of regional $A\beta$ accumulation throughout the brain are associated with APOE $\epsilon 4$ allele carriage. Finally, faster accumulation of $A\beta$ in nearly all regions was associated with elevated tau pathology, particularly in the temporal and parietal lobes, with little difference based on where $A\beta$ accumulates. These results indicate that $A\beta$ is unlikely to spread from one region to another, at least early in the disease process, but rather arises quickly across the brain perhaps based on shared molecular vulnerability. In addition, this study shows that those who have faster rates of $A\beta$ accumulation not only have elevated genetic risk for sporadic, late-onset AD, but also exhibit elevated burden of a second hallmark of AD, tau pathology.

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Disclosure statement

Molly R. LaPoint – conceptualization, methodology, software, formal analysis, data curation, writing-original draft, visualization; Suzanne L. Baker – software, methodology, data curation, writing-review and editing; Susan M. Landau – methodology, data curation, writing-review and editing; Theresa M. Harrison – software, writing-review and editing; William J. Jagust – conceptualization, methodology, writing-review and editing, supervision, funding acquisition. Suzanne L. Baker has consulted for Genentech. Susan M. Landau has served on the advisory board for KeifeRx. William J. Jagust has consulted for Biogen, Bioclinica and CuraSen. Molly R. LaPoint and Theresa M. Harrison have nothing to declare.

- We verify that the present manuscript has not been previously published or submitted to any other journal.
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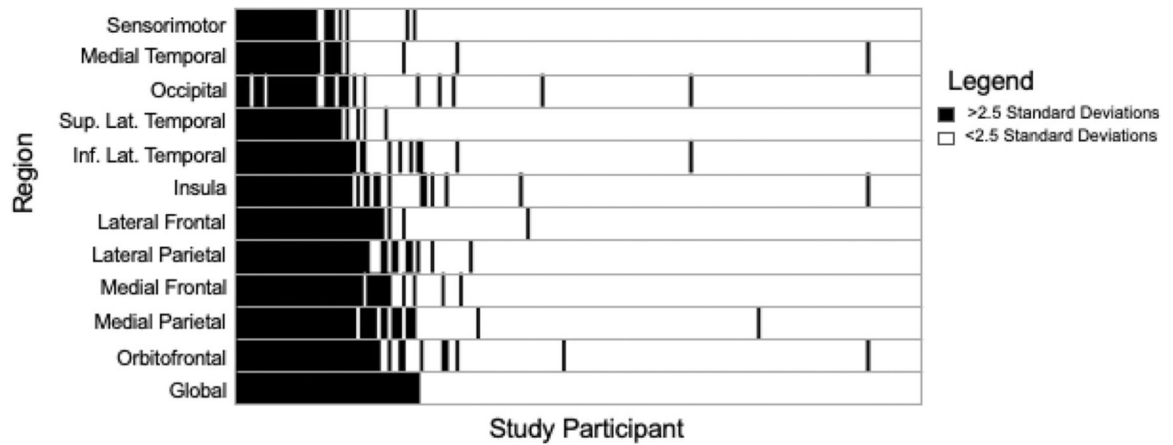


Fig. 1.

Regional PIB by subject using cross-sectional Z-score method. In this figure, each participant is represented on the x-axis, in descending order by their global PIB DVR, and the regions are listed with on the y-axis. For each participant, the regions with a Z-score above 2.5 standard deviations are filled in black. Most participants who have elevated PIB-PET in one region have it in most or all other regions.

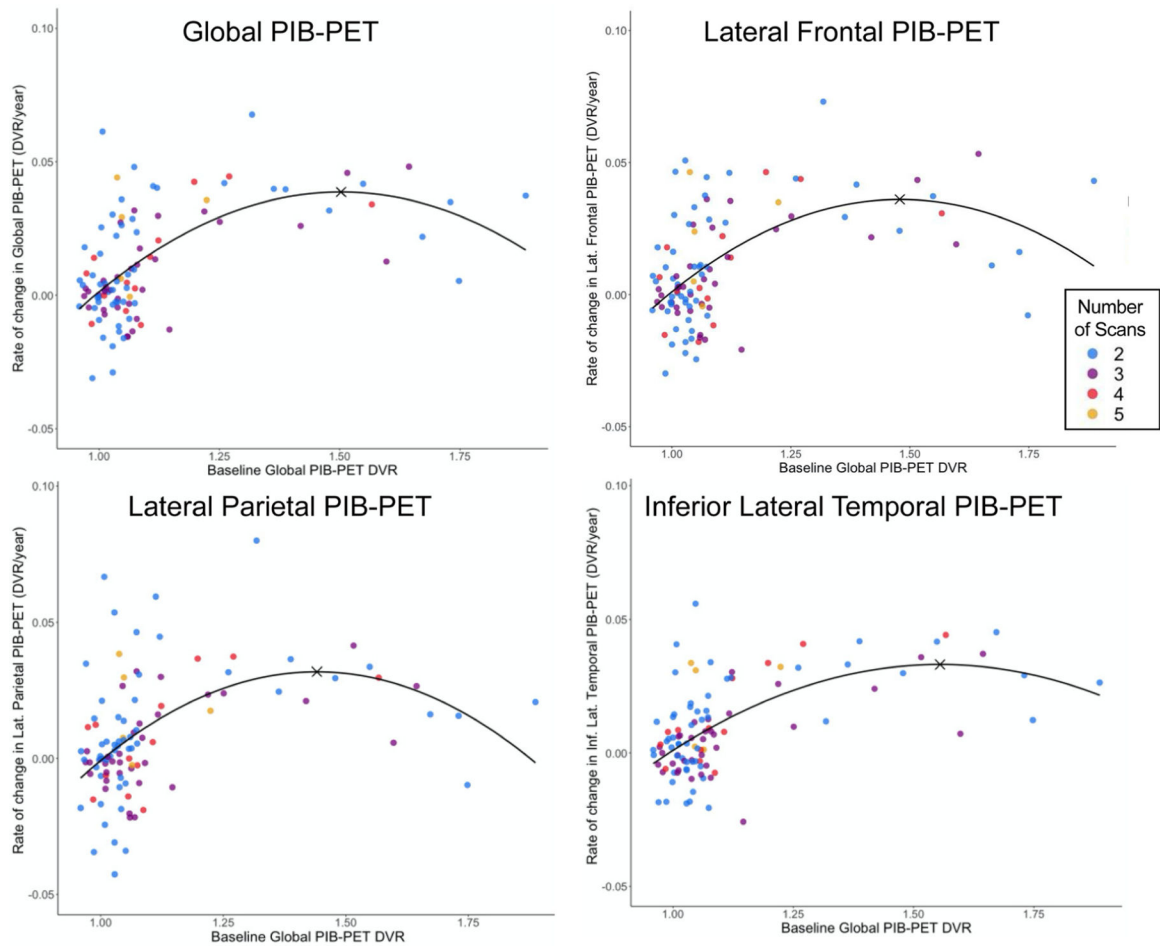


Fig. 2. Quadratic Fit for a subset of regions. These figures show the relationships between global PIB at baseline and regional accumulation of amyloid for a subset of the analyzed regions. Global PIB is on the x-axis and regional change (DVR/year) is on the y-axis. Study participants are color coded according to how many PIB scans they have received. The black “X” on each graph is the point at which the regional slope is at its maximum. After this time point, regional PIB begins to decelerate.

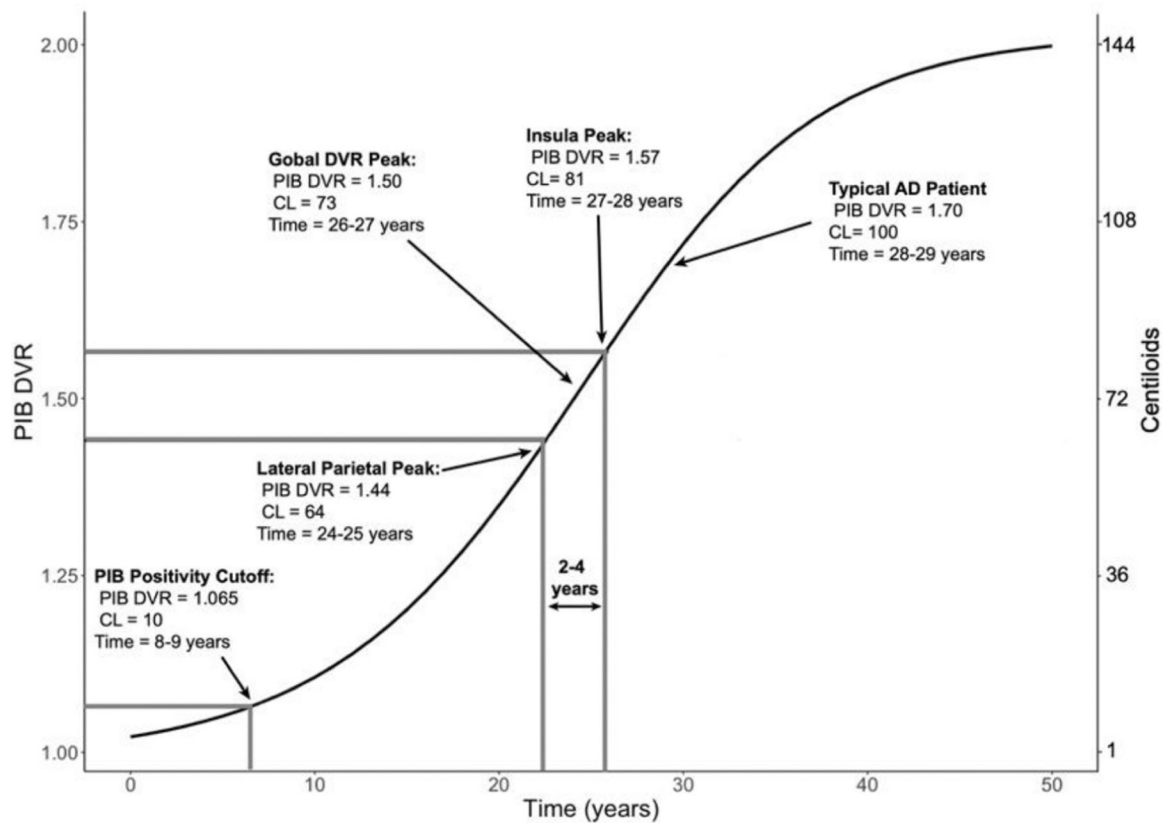


Fig. 3.

Sigmoidal time-global PIB relationship, showing different points of interest. This figure shows how global PIB DVR changes (y-axis) over 50 years, starting with the mean PIB DVR of amyloid-negative participants in the current study (DVR = 1.02). It takes 8–9 years to reach a DVR of 1.065, the cutoff for PIB positivity in this sample, and the 8 different regions reach their peak accumulation rates between 24 and 28 years from the mean baseline.

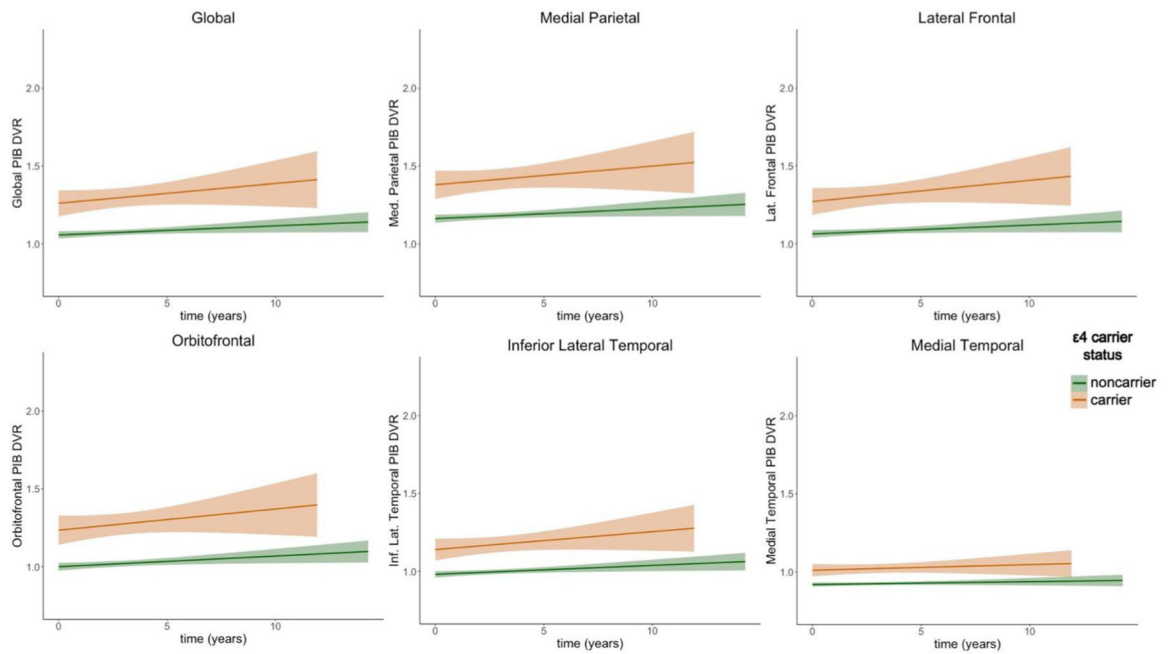


Fig. 4. PIB accumulation rates in $\epsilon 4$ carriers versus noncarriers. These plots show the accumulation over time in $\epsilon 4$ carriers versus noncarriers. $\epsilon 4$ carriers have significantly faster accumulation in all regions (see Table 4), controlling for baseline age and sex, and both terms interacting with time, in the linear mixed effects models. These figures are showing linear model fits in each group, not including any covariates.

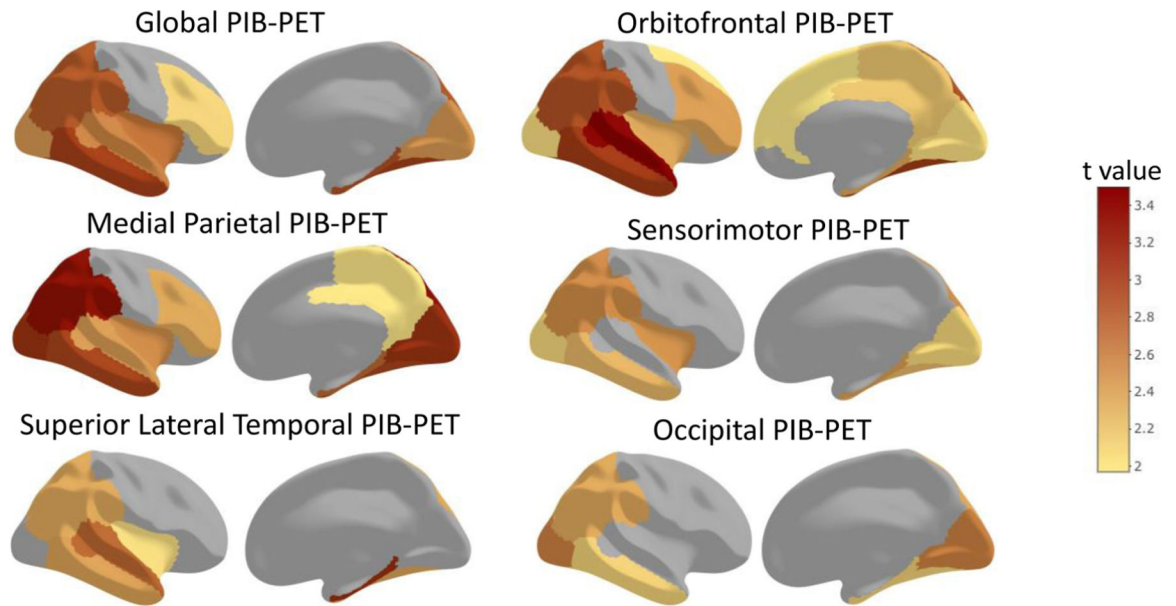


Fig. 5. PIB-Tau relationships. Colors reflect the association between regional $A\beta$ accumulation rates in the indicated brain region and regional FTP-PET at a liberal threshold of $p < 0.05$, without correction for multiple comparisons. The pattern of associations between $A\beta$ accumulation rates and tau deposition is similar across regions, with the strongest PIB-FTP relationships in temporal and lateral parietal regions and more moderate associations in global, OFC, and medial parietal PIB regions.

Subject demographics and descriptive statistics. A summary of all study participants. Participants in the longitudinal sample are a subset of the cross-sectional sample, and the FTP-PET sample is a further subset of the longitudinal sample

Table 1

Sample	Cross-sectional	Longitudinal	FTP-PET
N	195	106	88
Female/Male	112/83	66/40	54/34
Age (y)	75.5 ± 5.97	75.0 ± 5.13	74.6 ± 4.93
Y of Education	16.9 ± 2.33	16.8 ± 1.86	17.0 ± 2.54
Follow-up Time (y)	-	4.71 ± 3.07	4.91 ± 3.24
Number PIB scans (2,3,4,5)	-	52, 35, 13, 5	38, 32, 13, 5
PIB+/-	73/122	41/65	35/53
E4 carriers/noncarriers	45/134 (16 missing)	31/74 (1 missing)	25/62 (1 missing)

Table 2

Z-score method results. Cross-sectional analyses ranked regions based on the number of participants with elevated $A\beta$ in each region. Elevated $A\beta$ was defined as having a Z-score value above 2.5 standard deviations, based on the $A\beta$ -negative participants.

Region	Percent Involvement	Z-score mean
Global PIB	26.7%	2.87
OFC	25.6%	2.16
Medial Parietal	25.1%	2.12
Medial Frontal	24.1%	2.16
Lateral Parietal	23.1%	1.85
Lateral Frontal	23.1%	2.43
Insula	23.1%	1.60
Inferior Lateral Temporal	22.1%	2.21
Superior Lateral Temporal	17.4%	1.59
Occipital	17.4%	1.01
MTL	16.9%	0.98
Sensorimotor	15.4%	1.08

Table 3

Global PIB value at which each region reaches its peak accumulation rate. This table shows the global PIB DVR, Mean regional slope calculated with linear regression, Global PIB Centiloid value at peak, and number of years it takes for each region to reach its peak rate of accumulation using the quadratic fit. Parietal and frontal regions precede lateral temporal regions as well as the insula

Region	Global PIB DVR at peak regional slope	Mean slope (DVR/y)	Centiloid value at peak regional slope	Y to Peak Accumulation Rate
Lateral Parietal	1.44	0.008	64	24–25
Sensorimotor	1.46	0.007	67	25–26
Medial Parietal	1.47	0.011	68	25–26
Medial Frontal	1.48	0.013	69	25–26
Lateral Frontal	1.48	0.010	69	25–26
Global PIB	1.50	0.011	73	26–27
Inferior Lateral Temporal	1.56	0.009	80	27–28
Superior Lateral Temporal	1.56	0.007	80	27–28
Insula	1.57	0.006	81	27–28

Table 4

Results from $\epsilon 4$ LME analyses. T and p values from the linear mixed effects models run comparing amyloid accumulation in $\epsilon 4$ carriers versus noncarriers. Included in the model were age and baseline time, both interacting with time, as covariates (not shown above). The $\epsilon 4 \times$ time term represents the difference over time in the trajectories of the 2 groups, and the $\epsilon 4$ term is the overall difference between $\epsilon 4$ carriers and noncarriers. The analyses were coded such that the direction of the results indicates that $\epsilon 4$ carriers had higher values and faster accumulation than the noncarriers

Region	$\epsilon 4 \times$ time term t value	$\epsilon 4 \times$ time term p value	$\epsilon 4$ term t value	$\epsilon 4$ term p value
Global PIB	4.86	$p < 0.0001$	5.41	$p < 0.0001$
OFC	4.16	$p < 0.0001$	5.79	$p < 0.0001$
Medial Parietal	3.97	$p = 0.0001$	5.06	$p < 0.0001$
Medial Frontal	4.10	$p = 0.0001$	5.10	$p < 0.0001$
Lateral Parietal	4.18	$p < 0.0001$	5.31	$p < 0.0001$
Lateral Frontal	5.10	$p < 0.0001$	5.43	$p < 0.0001$
Insula	4.52	$p < 0.0001$	4.80	$p < 0.0001$
Inferior Lateral Temporal	4.76	$p < 0.0001$	4.81	$p < 0.0001$
Superior Lateral Temporal	4.19	$p < 0.0001$	4.47	$p < 0.0001$
Occipital	3.93	$p = 0.0007$	3.93	$p = 0.002$
MTL	3.90	$p = 0.0001$	4.21	$p = 0.0001$
Sensorimotor	3.17	$p = 0.002$	3.93	$p = 0.0002$

Table 5

Comparison of papers modeling the time course of amyloid accumulation. A comparison of 4 papers (the present study and 3 others) that have modeled the time course of amyloid accumulation in the ADNI, Alzheimer's disease neuroimaging initiative; MCSA, the mayo clinic study of aging; and the AIBL, Australian imaging biomarkers, and lifestyle study.

Study	This paper	ADNI (Jagust & Landau 2021)	MCSA (Jack et al. 2013)	AIBL (Villemagne et al. 2013)
Sigmoid starting value	4 CL (AB- average)	4 CL (AB- average)	1.3 PIB SUVR	1.17 PIB SUVR (AB- average for accumulators)
Time from baseline to AB+ threshold	8-9 y(25 CL)	7 y(25.3 CL)	6.63 y (SUVR 1.5 to peak)	12 y
Time from AB+ threshold to peak	17-18 y	3 y	-	~17 y ^a
Time from peak to typical AD patient	3-4 y(94 CL)	18 y(100 CL)	7.66 y (peak to SUVR 2.5)	~2 y ^a

^aEstimated from the papers' figures