

Longitudinal relationships among biomarkers for Alzheimer disease in the Adult Children Study

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ABSTRACT

Objective: To determine whether and how longitudinal rates of change in MRI volumetrics, CSF concentrations of Alzheimer-related proteins, molecular imaging of cerebral fibrillar amyloid with PET using the [^{11}C] benzothiazole tracer, Pittsburgh compound B (PiB), and cognition were associated among asymptomatic middle-aged to older individuals.

Methods: Multivariate mixed models for repeated measures were used to assess the correlations on the rates of changes across markers.

Results: Among 209 asymptomatic middle-aged to older individuals longitudinally followed for up to 11 years (mean 6.7 years), a faster intraindividual decrease in CSF $\text{A}\beta_{42}$ was associated with a faster increase in PiB mean cortical standardized uptake value ratio (MCSUVR, $p = 0.04$), but not others. The rate of change in CSF tau (and Ptau_{181}) was correlated with the rate of change in PiB MCSUVR ($p = 0.002$), hippocampal volume ($p = 0.04$), and global cognition ($p = 0.008$). The rate of change in hippocampal volume was correlated with the rate of change in global cognition ($p = 0.04$). Only 3 significant correlations were observed at baseline: CSF $\text{A}\beta_{42}$ and PiB MCSUVR ($p < 0.001$), CSF tau and PiB MCSUVR ($p < 0.001$), and CSF $\text{A}\beta_{42}$ and global cognition ($p = 0.01$).

Conclusions: CSF tau (Ptau_{181}), PiB MCSUVR, and hippocampal volume were all longitudinally correlated with each other, whereas CSF $\text{A}\beta_{42}$ was correlated only with PiB binding. Unlike the baseline values, the longitudinal change in CSF tau (Ptau_{181}) and hippocampal volume were correlated with the longitudinal change in global cognition, validating the role of these biomarkers in Alzheimer disease prevention trials. **Neurology**® 2016;86:1499-1506

GLOSSARY

A4 = Anti-Amyloid Treatment in Asymptomatic Alzheimer's; **A β 42** = the 42 amino acid isoform of the amyloid- β peptide; **ACS** = Adult Children Study; **AD** = Alzheimer disease; **BMMRM** = bivariate mixed model for repeated measures; **CDR** = Clinical Dementia Rating; **CI** = confidence interval; **DIAN TU** = Dominantly Inherited Alzheimer Network Trials Unit; **FH** = family history; **LP** = lumbar puncture; **MCSUVR** = mean cortical standardized uptake value ratio; **NFT** = neurofibrillary tangle; **PiB** = Pittsburgh compound B; **ROI** = region of interest; **SUVR** = standardized uptake value ratio; **WMS** = Wechsler Memory Scale.

Accumulating research suggests that neurodegenerative processes associated with Alzheimer disease (AD) may begin at middle age (~ 50 years)¹⁻³ and almost certainly many years prior to symptom onset.⁴⁻⁷ Because by definition there are no clinical symptoms at this preclinical stage of AD, biomarkers can be an effective tool to measure disease progression so that early interventions can be tested and developed. The cross-sectional and longitudinal associations across multiple modalities of AD biomarkers have been well-characterized, especially in elderly individuals 65 years or older with and without clinical symptoms of AD.⁸⁻¹⁰ It remains unknown, however, how and to what degree the longitudinal changes of AD biomarkers are correlated in asymptomatic middle-aged to older individuals. The objective of this report is to provide a comprehensive assessment of the biomarker relationships on the longitudinal rates of change across major modalities of AD biomarkers and cognition in a cognitively normal middle-aged to older cohort, and compare results to the cross-sectional correlations of baseline values alone.

Supplemental data
at Neurology.org

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METHODS Participants. Since 2005, the Adult Children Study (ACS) has enrolled a cohort of cognitively normal 43- to 77-year-old individuals in a comprehensive study of biomarkers for AD prior to its symptomatic stages.¹¹ In addition to clinical and cognitive measures, a broad spectrum of biomarkers for AD were longitudinally assessed, including MRI-based regional brain volumes, CSF analytes, and molecular imaging of cerebral fibrillar amyloid with PET using the [¹¹C] benzothiazole tracer, Pittsburgh compound B (PiB). As of October 2014, the ACS enrolled 357 community-living volunteers from the greater St. Louis metropolitan area. Recruitment primarily was through word of mouth and personal inquiries. The design of the ACS was a 2-way stratification by family history (FH) for late-onset AD and 3 age groups at baseline (43–54, 55–64, 65–77 years).¹² Eligibility criteria for the ACS were availability of an informant who knew the participant well, cognitive normality at baseline (defined as Clinical Dementia Rating [CDR]¹³ = 0), and willingness in principle to complete all study procedures at baseline and each follow-up. Individuals with comorbid conditions, including depressive features short of major affective disorder, were allowed in ACS if clinically stable at time of enrollment. Exclusion criteria included conditions that would preclude longitudinal participation or confound cognitive assessment or membership in families with a dominantly inherited pattern of AD or a known causative mutation for AD.

Standard protocol approvals, registrations, and patient consents. The Washington University Human Research Protection Office approved the study and all participants gave written informed consent.

Clinical and cognitive assessments. The clinical and cognitive assessments were conducted longitudinally every 3 years except for participants age 65 years or older, who were assessed annually. Details of clinical and cognitive assessments have been described previously.¹² In brief, the primary clinical assessment protocol is that of the National Alzheimer Coordinating Center Uniform Data Set,¹⁴ which includes standard definitions and diagnostic criteria for detection of dementia and its differential diagnosis.^{14,15} The presence or absence of dementia and, when present, its severity were operationalized with the CDR.¹³ The entire clinical assessment takes approximately 90 minutes to complete. Participants completed comprehensive psychometric testing 1–2 weeks after their clinical assessment. The 5 cognitive domains assessed in the 2-hour cognitive battery include episodic memory, working memory, semantic knowledge, executive function and attention, and visuospatial ability. A global cognition score covering all major cognitive domains for early changes was computed by using 7 cognitive tests that were shared by all age groups of the ACS cohort: Logical Memory Delayed and Verbal Paired Associates from the Wechsler Memory Scale (WMS),¹⁶ Free and Cued Selective Reminding,¹⁷ WMS-III Letter-Number Sequencing,¹⁸ Animal Naming,¹⁹ and Trailmaking Test A and B.²⁰ The global cognitive score represented the average of the *z* scores from all 7 tests, each of which was obtained by first subtracting the baseline mean score over the entire ACS cohort from each individual's score and then dividing the difference by the baseline SD.

CSF collection and analysis. Longitudinal CSF was collected in the ACS. The assessment protocol has been described previously.⁸ Briefly, CSF (20–30 mL) was collected by routine lumbar puncture (LP) in polypropylene tubes at 8:00 AM after overnight fasting as previously described.⁸ The samples were analyzed for total tau, tau phosphorylated at threonine-181 (Ptau₁₈₁), and Aβ1-42 (Aβ₄₂) by commercial ELISA (Improved INNOTEST,

Fujirebio, Ghent, Belgium).²¹ Longitudinal CSF samples from the same individual were run on the same assay plate (and same lot number) in order to minimize potential interplate and inter-lot methodologic variability. Samples were continuously kept on ice and assays were performed on the same aliquot after a single thaw following initial freezing.

Image acquisition and processing. MRI scans were obtained on a Sonata 1.5T, Vision 1.5T, or Trio 3.0T scanner (Siemens, Munich, Germany). All participants with longitudinal MRI assessments were included in this report, and all longitudinal scans were obtained on a Trio 3.0T scanner. Structural MRI processing steps have been described in detail previously^{12,22} and included motion correction, averaging across scans, and atlas transformation. Regional volumes were obtained via the FreeSurfer image analysis suite (Athinoula A. Martinos Center for Biomedical Imaging, Charlestown, MA). Hippocampal volume was selected as the region of interest (ROI) in this analysis.

PET PiB imaging and analysis procedures have been reported elsewhere.^{12,23} To summarize, PiB amyloid deposition in specified brain ROI was determined using FreeSurfer,^{24–26} and a standardized uptake value ratio (SUVR) with and without correction for partial volume effects^{27,28} was calculated for each ROI, based upon the last 30 minutes acquired as part of a 60-minute dynamic acquisition. Partial volume correction was performed using a regional spread function technique.²⁸ The mean cortical SUVR (MCSUVR) was calculated from FreeSurfer regions within the prefrontal cortex, precuneus, and temporal cortex. The cerebellum was chosen as the reference region.

Genotyping. *APOE* genotyping was performed as previously described.^{12,29}

Statistical analysis. All 209 individuals with available longitudinal data from at least 2 modalities (CSF, PET/PiB, MRI, and

Table 1 Demographic and biomarker descriptive statistics at baseline (n = 209)

Variables	Statistics
Age, y, mean (SD)	60.5 (8.4)
Female, n (%)	142 (67.9)
Family history positive, n (%)	119 (56.9)
<i>APOE4</i> positive, n (%)	78 (37.3)
Education, y, mean (SD)	16.2 (2.6)
MMSE, mean (SD)	29.3 (1.0)
CSF Aβ ₄₂ , pg/mL, mean (SD)	1,220.9 (352.5)
CSF tau, pg/mL, mean (SD)	278.5 (139.8)
Ptau ₁₈₁ , pg/mL, mean (SD)	51.7 (23.6)
PiB MCSUVR, mean (SD)	1.153 (0.231)
Positive PiB MCSUVR, n (%)	26 (13.5)
Hippocampal volume, mean (SD)	3,890.8 (438.0)
Global cognition, mean (SD)	0.014 (0.520)

Abbreviations: Aβ₄₂ = the 42 amino acid isoform of the amyloid-β peptide; MCSUVR = mean cortical standardized uptake value ratio; MMSE = Mini-Mental State Examination; PiB = Pittsburgh compound B. Positive PiB MCSUVR was determined using a threshold of 1.31.

cognition) were analyzed. For each pair of markers, a bivariate mixed model for repeated measures (BMMRM) was used to simultaneously model the longitudinal courses of both markers.^{30,31} Specifically, random intercept and slope (i.e., the rate of change) were assumed for each marker across individuals.³² The entire set of random effects from 2 markers including 2 slopes and 2 intercepts from the BMMRM was then assumed to follow a joint 4D multivariate normal distribution with an unstructured covariance matrix across the participants. The unstructured covariance matrix is important because it allows an unbiased assessment on the correlation of the rates of change across biomarkers. Different residual variances were assumed between markers. All fixed effects and variance/covariance components were estimated by the method of maximum likelihood. BMMRM was chosen because of the concern on the convergence in the maximum likelihood estimation with more than 2 markers in the joint analyses. A 95% confidence interval (CI) for the correlation of the bivariate rates of change was estimated through the Delta method after the Fisher *z* transformation. To adjust for the effect of baseline age, FH, and *APOE* $\epsilon 4$ allele (*APOE4*) status, similar BMMRMs were fitted by including the fixed effects of these covariates in both the slopes and the intercepts. These models were flexible in handling unbalanced and unevenly spaced multivariate longitudinal biomarker data as well as missing data in the ACS, and have been previously used in studies of AD.³³ Model diagnostics (i.e., residual plots) indicated no major concern on the model goodness-of-fit. All models were implemented in PROC MIXED/SAS.³⁴

RESULTS Table 1 presents the baseline demographics and biomarkers of 209 individuals in the ACS cohort who were part of the analyses. A total of 145 (69%) individuals were younger than 65 or age 65 years at baseline. The duration and the frequency of longitudinal follow-up for each modality are given in table 2. A total of 207/209 individuals underwent longitudinal clinical and cognitive assessments for up to 11 years. Fifteen progressed to higher CDR after baseline. A total of 169 (81%) had longitudinal LPs to obtain CSF, 152 (73%) completed longitudinal PET PiB scans, and 164 (78%) underwent longitudinal MRI scans.

Table 2 Summary of longitudinal follow-up in the Adult Children Study cohort

Variables	Statistics
CSF: length of follow-up, y, mean (SD) (range)	4.6 (1.7) (1.7–9.3)
No. serial CSF observations (2/3/4)	102/62/5
PET PiB: length of follow-up, y, mean (SD) (range)	4.7 (1.8) (1.1–9.2)
No. serial PiB observations (2/3/4/5)	90/58/3/1
MRI: length of follow-up, y, mean (SD) (range)	4.2 (1.5) (0.9–6.5)
No. serial MRI observations (2/3/4)	103/55/6
Cognition: length of follow-up, y, mean (SD) (range)	6.7 (2.0) (3.0–10.7)
No. serial cognitive assessments (2/3/4/≥5)	30/47/35/95

Abbreviation: PiB = Pittsburgh compound B.

Number of serial assessments is presented in the format of "r/s/t/u," where r is the number of subjects with 2 serial assessments, s is the number of subjects with 3 serial assessments, t is the number of subjects with 4 serial assessments, u is the number of subjects with at least 5 serial assessments.

Table 3 presents the associations among CSF $A\beta_{42}$, CSF tau (and Ptau₁₈₁), PiB MCSUVR, hippocampal volume, and global cognition at baseline alone. The dashed lines between 2 biomarkers in figure 1 represent statistically significant correlations at baseline ($p < 0.05$), whereas biomarkers without connecting dashed lines are not significantly correlated at baseline. Specifically, after adjusting for the effects of family history, baseline age, and *APOE4* status, only 3 significant correlations were observed at baseline: CSF $A\beta_{42}$ and PiB MCSUVR ($r = -0.49$, 95% CI -0.63 to 0.32), CSF tau and PiB MCSUVR ($r = 0.52$, 95% CI 0.36 – 0.65), and CSF $A\beta_{42}$ and global cognition ($r = 0.22$, 95% CI 0.05 – 0.39).

Table 4 presents the estimated correlations on the longitudinal rates of change for the same biomarker and cognitive measures (see tables e-1 through e-4 on the *Neurology*[®] Web site at Neurology.org for the estimated rates of changes and figures e-1 and e-2 for plots). Results for MCSUVR with and without correction for partial volume effects²⁸ were very similar, so only those without correction are presented. The solid lines between 2 biomarkers in figure 1 represent statistically significant correlations on the rates of change ($p < 0.05$), whereas biomarkers without solid line connections are not significantly correlated in their longitudinal changes. The rate of change in CSF $A\beta_{42}$ was correlated with the rate of change in PiB MCSUVR ($r = -0.40$, 95% CI -0.67 to -0.03), indicating that among asymptomatic middle-aged to older individuals, a faster decrease of CSF $A\beta_{42}$ over time is associated with a faster increase of PiB MCSUVR. The rate of change in CSF $A\beta_{42}$ was not significantly correlated with that of CSF tau (or Ptau₁₈₁), hippocampal volume, or global cognition. Longitudinal rate of change in CSF tau, however, was correlated with longitudinal rate of change in PiB MCSUVR ($r = 0.53$, 95% CI 0.27 – 0.71), hippocampal volume ($r = -0.51$, 95% CI -0.74 to -0.18), and global cognition ($r = -0.50$, 95% CI -0.71 to -0.21). Similar significant correlations were also observed between the longitudinal rate of change in CSF Ptau₁₈₁ and that in PiB MCSUVR, hippocampal volume, and global cognition. Further, the rate of change in hippocampal volume was not significantly correlated with that in PiB MCSUVR ($r = -0.22$, 95% CI -0.49 to 0.08), but was correlated with the rate of change in global cognition ($r = 0.49$, 95% CI 0.09 – 0.75). In contrast, the rates of change in PiB MCSUVR and global cognition were not significantly correlated ($r = -0.24$, 95% CI -0.48 to 0.03). The correlations of biomarkers on the rate of change after adjusting for the effects of FH, baseline age, and *APOE4* were consistent with those from the unadjusted analyses (table 4). Additional sensitivity analyses by further adjusting for the effect of sex, education,

Table 3 Biomarker correlations (95% confidence interval) and p value on the baseline values, without (the first 4 numbers of each cell) and with (the last 4 numbers of each cell) adjusting for the effects of APOE4, family history, and baseline age

	CSF tau	CSF Ptau ₁₈₁	PIB MCSUVR	Hippocampal volume	Global cognition
CSF Aβ₄₂	0.06 (-0.11 to 0.22), p = 0.51; 0.14 (-0.03 to 0.30), p = 0.12	0.06 (-0.10 to 0.22), p = 0.47; 0.10 (-0.07 to 0.26), p = 0.24	-0.55 (-0.67 to -0.41), p < 0.001; -0.49 (-0.63 to -0.32), p < 0.001	-0.04 (-0.23 to 0.16), p = 0.71; -0.13 (-0.33 to 0.08), p = 0.22	0.25 (0.08 to 0.41), p = 0.004; 0.22 (0.05 to 0.39), p = 0.01
CSF tau			0.59 (0.45 to 0.70), p < 0.001; 0.52 (0.36 to 0.65), p < 0.001	-0.23 (-0.40 to -0.04), p = 0.02; -0.12 (-0.31 to 0.09), p = 0.26	-0.09 (-0.26 to 0.08), p = 0.31; -0.03 (-0.20 to 0.15), p = 0.76
CSF Ptau₁₈₁			0.54 (0.39 to 0.66), p < 0.001; 0.48 (0.32 to 0.62), p < 0.001	-0.22 (-0.39 to -0.03), p = 0.02; -0.13 (-0.32 to 0.07), p = 0.21	-0.11 (-0.28 to 0.06), p = 0.20; -0.06 (-0.24 to 0.11), p = 0.48
PIB MCSUVR				-0.23 (-0.40 to -0.06), p = 0.009; -0.15 (-0.32 to 0.04), p = 0.12	-0.24 (-0.40 to -0.07), p = 0.006; -0.15 (-0.33 to 0.03), p = 0.10
Hippocampal volume					0.14 (-0.03 to 0.31), p = 0.11; 0.11 (-0.08 to 0.29), p = 0.25

Abbreviations: Aβ₄₂ = the 42 amino acid isoform of the amyloid-β peptide; BMMRM = bivariate mixed model for repeated measures; MCSUVR = mean cortical standardized uptake ratio; PIB = Pittsburgh compound B.

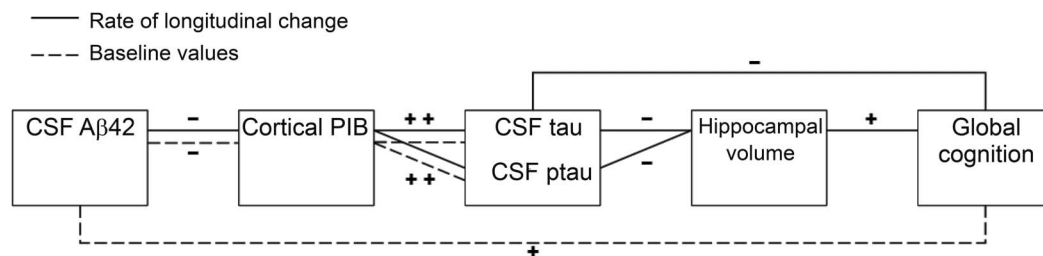
The p value in each cell is for testing the null hypothesis that the correlation equals 0 in the BMMRM.

hypertension, diabetes, medications, and the length of follow-up as well as by excluding subjects who were biomarker positive at baseline or progressed to higher CDR resulted in largely consistent estimates of the biomarker correlations on the rate of longitudinal change. Finally, baseline values of CSF Aβ₄₂, CSF tau, Ptau₁₈₁, PiB MCSUVR, and hippocampal volume were all significantly associated with the longitudinal rate of global cognition (table e-2).

DISCUSSION Clinicopathologic studies demonstrate that asymptomatic elderly individuals can manifest the neuropathologic changes of AD, notably senile plaques and neurofibrillary tangles (NFT).^{2,4,35} Such neuropathologic changes may begin in middle age.¹⁻³ In the absence of clinical and cognitive symptoms, biomarkers can be an effective tool to measure disease progression so that early interventions can be tested and developed. Ongoing secondary prevention trials including the Anti-Amyloid Treatment in Asymptomatic Alzheimer's (A4) trial, the Dominantly Inherited Alzheimer Network Trials Unit (DIAN TU) trials, and the Alzheimer's Prevention Initiative trial all depend on the ability to recruit asymptomatic individuals with the highest probability of manifesting measurable, reliable cognitive changes over a given study period. Understanding the longitudinal relationship among biomarkers and cognitive measures is of paramount importance.

The cross-sectional associations across established AD biomarkers have been well-characterized, especially in elderly individuals 65 years or older with and without clinical symptoms of AD.⁸⁻¹⁰ It remains unknown, however, how and to what degree the longitudinal changes of these biomarkers are correlated in asymptomatic middle-aged to older individuals. To our knowledge, our report represents the first attempt to address this question with all major AD biomarker modalities. Our findings reveal that the longitudinal rates of change in different modalities of markers and cognition are already correlated among asymptomatic middle-aged to older individuals. Specifically, the rates of change in CSF biomarkers (Aβ₄₂, tau, and Ptau₁₈₁) were all correlated with that of PET PiB MCSUVR, which was then correlated with the rate of change in hippocampal volume. Furthermore, the rates of change in CSF tau, Ptau₁₈₁, and MRI hippocampal volume were also correlated with the rate of cognitive decline. These correlations establish the longitudinal validity of CSF tau and Ptau₁₈₁ and MRI volumetrics as prognostic biomarkers for cognitive decline in preclinical AD, and also imply the validity of PET PiB MCSUVR and CSF Aβ₄₂ in tracking early disease progression years prior to symptomatic onset. Thus, these findings support the current design of ongoing secondary prevention trials of AD in

Figure 1 Graphical presentation of longitudinal and cross-sectional (i.e., at baseline) biomarkers correlations across multiple modalities



Correlations are after adjusting for the effects of *APOE4*, family history, and baseline age. Solid lines identify significant longitudinal correlations on the rates of change, and dashed lines significant cross-sectional correlations at baseline. Markers that do not significantly correlate are not connected by lines; + and - represent positive and negative correlations, respectively. PiB = Pittsburgh compound B.

which biomarkers are either the primary inclusion and exclusion criteria (the A4 trial) or the primary efficacy outcomes (DIAN TU) prior to a critical phase III trial that will utilize a cognitive endpoint.

Perhaps some of the most important findings from our study are the nonsignificant correlations from a relatively large cohort of asymptomatic middle-aged to older individuals. It has been well-established that senile plaques and NFT are the hallmark neuropathologies of AD. Essentially all brains from individuals who die with late-onset AD exhibit both pathologies at autopsy,³⁶ and significant pathologic overlap also exists among individuals who died with absence of clinical symptoms.⁵ Yet we found no significant correlations, either cross-sectionally or longitudinally, between CSF Aβ₄₂ and CSF tau (or Ptau₁₈₁) in the asymptomatic middle-aged to older individuals, suggesting statistically independent early Aβ- and tau-related pathologic processes during early disease stages, consistent with our prior models.³⁷ One possible explanation for this lack of relationship is differential time windows within individuals during which one neuropathologic process has started whereas the other process remains either latent or only fluctuated in a random fashion independent of the former process, especially after adjusting for the effect of major covariates (FH, baseline age, and *APOE4*). It is also important to note that the observed correlation on the adjusted rate of change between CSF Aβ₄₂ and CSF tau (or Ptau₁₈₁) is fairly high (0.4). Hence the lack of relationship may be due to limited statistical power. Although the rate of change in CSF Aβ₄₂ was not correlated with the rate of cognitive decline, the baseline values of CSF Aβ₄₂ (table e-2), along with the rates of changes in CSF tau and Ptau₁₈₁, were already correlated with that of cognition in the middle-aged to older asymptomatic individuals, suggesting that the changes in CSF tau and Ptau₁₈₁ as well as the baseline levels of CSF Aβ₄₂ all predict the changes in global cognition.

Further, the rate of change in PiB MCSUVR is associated with the rate of change in all major CSF biomarkers including CSF Aβ₄₂, tau, and Ptau₁₈₁, but not in global cognition, supporting the argument that longitudinal changes in amyloid may precede that of cognition for a relatively long duration. Figure 1 demonstrates a pattern of longitudinally correlated biomarkers that is consistent with the hypothesized temporal orderings,³⁸ i.e., the adjacent biomarkers from the hypothesized biomarker temporal orderings were correlated on their rates of changes, whereas biomarkers farther distant and not adjacent to each other were not significantly correlated. Although our findings support the recently proposed diagnostic criteria of preclinical AD,³⁹ it is important to note that correlations on the rates of changes in biomarkers do not themselves indicate a temporal ordering of these markers.

In comparison, at baseline, the only significant correlations were between CSF Aβ₄₂ and PiB MCSUVR, CSF tau (or Ptau₁₈₁) and PiB MCSUVR, and CSF Aβ₄₂ and global cognition. Although baseline CSF tau, Ptau₁₈₁, and hippocampal volume were not correlated with baseline global cognition, their longitudinal rates of change were all significantly correlated with the rate of cognitive decline, long before the symptomatic onset of AD. Recently revised guidelines from the Food and Drug Administration for clinical trials in early-stage AD still mandate that treatments only be approved if they demonstrate cognitive and functional benefits. Our results suggest that, in order to adequately power future prevention trials of AD using cognitive outcomes, these trials may need to focus on individuals in the time window when their biomarker values in CSF tau, Ptau₁₈₁, and hippocampal volume are starting to change, because these changes predict changes in cognition. Intriguingly, at baseline, CSF Aβ₄₂ and global cognition were correlated, but their rates of longitudinal change were not. These observations suggest that, whereas

Table 4 Biomarker correlations (95% confidence interval) and p value on the longitudinal rate of change per year, without (the first 4 numbers of each cell) and with (the last 4 numbers of each cell) adjusting for the effects of APOE4, family history, and baseline age

	CSF tau	CSF Ptau ₁₈₁	PIB MCSUVR	Hippocampal volume	Global cognition
CSF Aβ₄₂	0.21 (-0.20 to 0.56), p = 0.31; 0.40 (-0.01 to 0.70), p = 0.06	0.24 (-0.16 to 0.58), p = 0.24; 0.40 (-0.01 to 0.69), p = 0.06	-0.40 (-0.67 to -0.03), p = 0.03; -0.39 (-0.67 to -0.01), p = 0.04	-0.19 (-0.63 to 0.35), p = 0.49; -0.34 (-0.77 to 0.31), p = 0.31	0.12 (-0.37 to 0.56), p = 0.64; -0.02 (-0.49 to 0.45), p = 0.93
CSF tau			0.53 (0.27 to 0.71), p < 0.001; 0.51 (0.21 to 0.72), p = 0.002	-0.51 (-0.74 to -0.18), p = 0.004; -0.41 (-0.69 to -0.02), p = 0.04	-0.50 (-0.71 to -0.21), p = 0.002; -0.44 (-0.68 to -0.12), p = 0.008
CSF Ptau₁₈₁			0.55 (0.31 to 0.73), p < 0.001; 0.56 (0.29 to 0.75), p < 0.001	-0.53 (-0.75 to -0.18), p = 0.004; -0.46 (-0.73 to -0.06), p = 0.03	-0.41 (-0.66 to -0.08), p = 0.02; -0.33 (-0.61 to 0.02), p = 0.07
PIB MCSUVR				-0.22 (-0.49 to 0.08), p = 0.15; -0.17 (-0.47 to 0.16), p = 0.31	-0.24 (-0.48 to 0.03), p = 0.09; -0.27 (-0.53 to 0.03), p = 0.07
Hippocampal volume					0.49 (0.09 to 0.75), p = 0.02; 0.47 (0.02 to 0.76), p = 0.04

Abbreviations: Aβ₄₂ = the 42 amino acid isoform of the amyloid-β peptide; BMMRM = bivariate mixed model for repeated measures; MCSUVR = mean cortical standardized uptake value ratio; PIB = Pittsburgh compound B.

The p value in each cell is for testing the null hypothesis that the correlation equals 0 in the BMMRM. Statistical significance is defined by p < 0.05. With a more stringent control of type I error rate due to possible multiplicity adjustment, table e-3 presents the sample sizes that will be needed to detect these correlations in table 4 with a statistical power of 80%.

CSF Aβ₄₂ may be associated with cognition across middle-aged to older asymptomatic individuals at a single time point, the intraindividual change of CSF Aβ₄₂ from baseline does not predict the intraindividual cognitive decline. Because these results are not entirely consistent with some of the previous reports,⁴⁰ future studies are needed to fully understand the biological and behavioral mechanisms behind this observation.

The major strengths of the study include the wide baseline age starting from ~43 years and large sample size of carefully characterized cognitively normal individuals for whom all major AD biomarkers were obtained longitudinally. The study also has limitations. First, the ACS is an observational study on a convenience sample. Unobserved factors could contribute to and confound the findings. Second, although the longitudinal follow-up was relatively long, it was not long enough to cover the entire preclinical disease course, thus preventing us from evaluating the cascade of early AD pathogenesis events in its entirety. It is also possible that there is a temporal lag in biomarker associations, which would require longitudinal change points or inflection points models over a relatively large number of longitudinal follow-ups. The ongoing longitudinal follow-up in the ACS cohort will provide much more insight to comprehensively understand the preclinical progression of AD.

AUTHOR CONTRIBUTIONS

C.X., A.M.F., T.L.S.B., D.H., J.H., V.B., K.L.M., and J.C.M. all contributed to the design of the study. D.H. and J.H. contributed to the cognitive data collection and analyses. A.M.F. and C.L.S. contributed to the collection of CSF samples and CSF data analyses. T.L.S.B. contributed to imaging data processing and analyses. C.X., M.S.J., and E.G. contributed to statistical analyses and manuscript drafting. E.G. contributed to database analyses. All authors contributed to the critical review of the manuscript.

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DISCLOSURE

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