

Testing the causal impact of amyloidosis on total Tau using a genetically informative sample of adult male twins.

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Statement of work

The manuscript is original research. It not been previously published and has not been submitted for publication elsewhere while under consideration.

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Abstract

INTRODUCTION: The amyloid cascade hypothesis predicts that amyloid-beta ($A\beta$) aggregation drives tau tangle accumulation. We tested competing causal and non-causal hypotheses regarding the direction of causation between $A\beta$ 40 and $A\beta$ 42 and total Tau (t-Tau) plasma biomarkers.

METHODS: Plasma $A\beta$ 40, $A\beta$ 42, t-Tau, and neurofilament light chain (NFL) were measured in 1,035 men (mean = 67.0 years) using Simoa immunoassays. Genetically informative twin modeling tested the direction of causation between $A\beta$ s and t-Tau.

RESULTS: No clear evidence that $A\beta$ 40 or $A\beta$ 42 directly causes changes in t-Tau was observed; the alternative causal hypotheses also fit the data well. In contrast, exploratory analyses suggested a causal impact of the $A\beta$ biomarkers on NFL. Separately, reciprocal causation was observed between t-Tau and NFL.

DISCUSSION: Plasma $A\beta$ 40 or $A\beta$ 42 do not appear to have a direct causal impact on t-Tau. In contrast, $A\beta$ aggregation may causally impact NFL in cognitively unimpaired men in their late 60s.

Keywords

Cascade hypothesis, amyloid-beta, tau, twin, gene, direction of causation

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Introduction

According to the amyloid cascade hypothesis [1], amyloid-beta ($A\beta$) aggregation drives accumulation of tau tangles, resulting in synaptic dysfunction, neurodegeneration and progression to cognitive decline. This implies a causal link from $A\beta$ aggregation to tau tangles. To our knowledge this link has not been empirically tested using genetically informative direction-of-causation modeling.

We previously explored the heritability of blood-based biomarkers related to risk of Alzheimer's Disease in a population-based sample of early old-age men [2]. Additive genetic influences explained 44% to 52% of the total variances in $A\beta_{42}$, $A\beta_{40}$, total tau (t-Tau), and neurofilament light chain (NFL), a marker of neurodegeneration. All remaining variances were explained by non-shared environmental influences. Since $A\beta$ aggregation was best explained by genetic and non-shared environmental influences [2], if either $A\beta$ phenotypically causes t-Tau, then significant genetic *and* environment covariance should be observed between $A\beta$ and t-Tau biomarkers. Instead, we found that both $A\beta_{42}$ and $A\beta_{40}$ were genetically uncorrelated with t-Tau. While this is consistent with there being no causal association between $A\beta$ and t-Tau, we apply a validated means [3] of empirically testing competing hypotheses regarding the direction of causation.

Aim

Without randomized control trials, Mendelian Randomization or longitudinal data, testing causality between complex traits is difficult. However, by analyzing genetically informative twin data and leveraging the expected differences in the patterns of cross-twin cross-trait correlations, it is possible to falsify hypotheses about the direction of causation between two variables measured on a single occasion [3-7]. Using this approach, we tested competing hypotheses regarding the direction of causation between $A\beta$ and t-Tau plasma biomarkers. We also included exploratory analyses modelling the direction of causation between $A\beta$ and NFL.

Methods

Subjects

For detailed sample and data description see Gillespie et al. [2]. The present study comprised of men from the Vietnam Era Twin Study of Aging (VETSA) who participated in a third assessment wave (mean age=68.2, SD=2.5, range=61.4 to 73.3) when plasma biomarkers were examined [2].

Blood-based biomarker data

Blood was collected under fasting conditions before acquisition and storage at -80°C . The Simoa Human Neurology 3-plex A (N3PA) Immunoassay was used to measure $A\beta_{40}$, $A\beta_{42}$, and t-Tau, while the Simoa NF-light assay was used to measure NFL (QuanterixTM, Billerica, MA, USA). Biomarkers were regressed onto age at assessment, testing site, storage time, self-reported race/ethnicity, and whether or not twins pairs were assessed on the same day. Residual scores were calculated using the `umx_residualize` function [8]. Next, the data were normal ranked in $R_{4.0.3}$ [9] and absolute values greater than three standard deviations (SDs) were eliminated to reduce

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skew. This eliminated a total of 12, 8, 12 and 15 subjects with A β 40, A β 42, NFL and t-Tau data respectively. Depending on the biomarker, there were between 988 to 1035 individuals (58% monozygotic and 42% dizygotic twins) with complete data.

Statistical analyses

Based on biometrical genetic methods [10], the OpenMx_{2.20.6} software package [11] with the raw data Full Information Maximum Likelihood (FIML) option and NPSOL optimizer in R_{4.2.2} [9] was used to decompose the total variance in each biomarker into additive genetic (A), shared or common (C) environment, and non-shared or unique (E) environmental influences while testing competing causal and non-causal hypotheses [10] (see Supplementary).

Figure 1 here

Illustrated in Figure 1, our null hypothesis predicted that associations between the A β and t-Tau biomarkers were explained by correlated, non-causal genetic and environmental influences (for brevity, only genetic influences are shown). We analyzed A β 40 and A β 42 separately and tested four competing, nested hypotheses: **(b)** A β causes t-Tau via β_1 ; **(c)** t-Tau causes A β via β_2 ; **(d)** reciprocal causation between A β and t-Tau via β_1 and β_2 ; and **(e)** no association i.e. $\beta_1 = \beta_2 = 0$. We also modelled the joint impact of both A β s on t-Tau (see Supplementary Figure S1), followed by exploratory causal modelling between the A β s and NFL, and finally between t-Tau and NFL.

The goodness of fit for each model was determined using the likelihood ratio statistic, which is the change in the minus two log-likelihood between the null and each competing model. This statistic, Δ -2LL, is asymptotically distributed as chi-squared with degrees of freedom equal to the difference in the number of free parameters between the null and each competing model. Our determination of the best-fitting model was also based on the optimal balance of complexity and explanatory power using Akaike's Information Criterion (AIC) [12].

Results

Model fit comparisons are shown in Table 1. For each set of analyses (i, ii & iii) the null hypothesis predicted that any observed association between A β and Tau was attributable to non-causal, correlated genetic and environmental factors.

When testing the 'A β 40 causes Tau' hypothesis, all four competing hypotheses (both uni-directional, the reciprocal, and the no association model) provided a good fit to the data in terms of non-significant changes in chi-square, whereas the 'no association' hypothesis provided the lowest AIC.

When testing the 'A β 42 causes Tau' hypothesis, the changes in chi-square for all three competing causal hypotheses were non-significant, whereas the AIC was lowest for the 'Tau causes A β 42' hypothesis. Note that the 'no association' hypothesis could be

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rejected based on the significant change in chi-square and higher AIC value relative to the three other competing hypotheses.

When testing the 'A β 40 & A β 42 (combined) cause Tau' hypothesis, the changes in chi-square for each of the three causal hypotheses were again non-significant. However, very little separated their corresponding AIC values. Note that the 'no association' hypothesis could again be rejected in terms of the significant chi-square change and highest AIC.

In exploratory analyses, we modelled the multivariate impact of A β 40 and A β 42 (A β s) on NFL (Supplementary Table S1). Among the competing hypotheses, the uni-directional 'A β s cause NFL' hypothesis provided a marginally better fit to the data as judged by the non-significant change in chi-square and lowest AIC. Finally, when modelling the association between t-Tau and NFL, the reciprocal causation model provided a (marginally) best fit to the data, followed next by the 't-Tau causes NFL' hypothesis.

Table 1 here

Discussion

To our knowledge, this is the first genetically informative test of the direction of causation between blood-based biomarkers related to Alzheimer's Disease. We found no unequivocal support for a causal impact of either A β 40 or A β 42 on t-Tau. Instead, alternative uni-directional and reciprocal hypotheses provided comparable fits to the data. In contrast, exploratory analyses suggest a causal impact of both blood-based A β biomarkers on NFL, and a reciprocal causal association between t-Tau and NFL.

The absence of clear, empirical support for a causal impact of A β on t-Tau, which would be consistent with the amyloid-beta cascade hypothesis, should be interpreted in the context of four considerations. First, our sample was predominately cognitively unimpaired. The proportion of men with mild cognitive impairment (MCI) was 15%. Causal signals may emerge as the sample ages and the prevalence of MCI increases over time. Second, we relied on plasma biomarkers. While accessible, affordable, and heritable [2], we note that dilution, degradation, and metabolism may introduce variation unrelated to AD-related brain changes. This may limit the predictive validity of these plasma biomarkers to model causation. Ultrasensitive immunoassays and novel mass spectrometry techniques that attempt to address this limitation have begun to show promise in terms of better plasma biomarker measurement [13, 14]. These two limitations are underscored by Coomans et al. [15] who analyzed data from a very small sample of older monozygotic twins with a relatively large number of APOE- ϵ 4 carriers and found significant associations between A β -PET and tau-PET. Third, to the extent that plasma A β is brain derived, it may nevertheless reflect general health conditions rather than brain amyloid accumulation. Finally, we relied on total tau rather than phosphorylated Tau (p-Tau), which aggregates into neurofibrillary tangles and is therefore likely to be a more relevant indicator of AD pathogenic processes. Indeed, the p-Tau 181, 217 and 231 isoforms have been shown to predict amyloidosis and

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progression to AD [60]. The genetic variance of these isoforms remains undetermined (including their covariance and direction of causation) with the A β and NFL biomarkers. Unlike our results for t-Tau, it is plausible that direction of causation modeling with p-Tau might be consistent with the amyloid cascade hypothesis.

Conclusion

Notwithstanding the absence of a population-based same-age replication sample, to the extent that plasma biomarkers are considered informative peripheral indicators of prodromal AD [16-18], our analyses suggest that neither A β 40 or A β 42 has any causal impact on t-Tau, when based on community-dwelling sample of men in their late 60s.

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Acknowledgements

The U.S. Department of Veterans Affairs, Department of Defense, National Personnel Records Center, National Archives and Records Administration, Internal Revenue Service, National Opinion Research Center, National Research Council, National Academy of Sciences, Institute for Survey Research, and Temple University provided invaluable assistance in the conduct of the VET Registry. The Cooperative Studies Program of the U.S. Department of Veterans Affairs provided financial support for development and maintenance of the Vietnam Era Twin Registry. We would also like to acknowledge the continued cooperation and participation of the members of the VET Registry and their families

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Conflicts

No authors reported a conflict of interest.

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Funding

This work was supported by the National Institute on Aging at the National Institutes of Health grant numbers R01s AG050595, AG022381, AG037985; R25 AG043364, F31 AG064834; P01 AG055367, AG062483; and K01 AG063805. The funding sources had no role in the preparation, review, or approval of the manuscript, or the decision to submit the manuscript for publication. The content of this manuscript is solely the responsibility of the authors and does not necessarily represent the official views of the NIA/NIH, or the VA.

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References

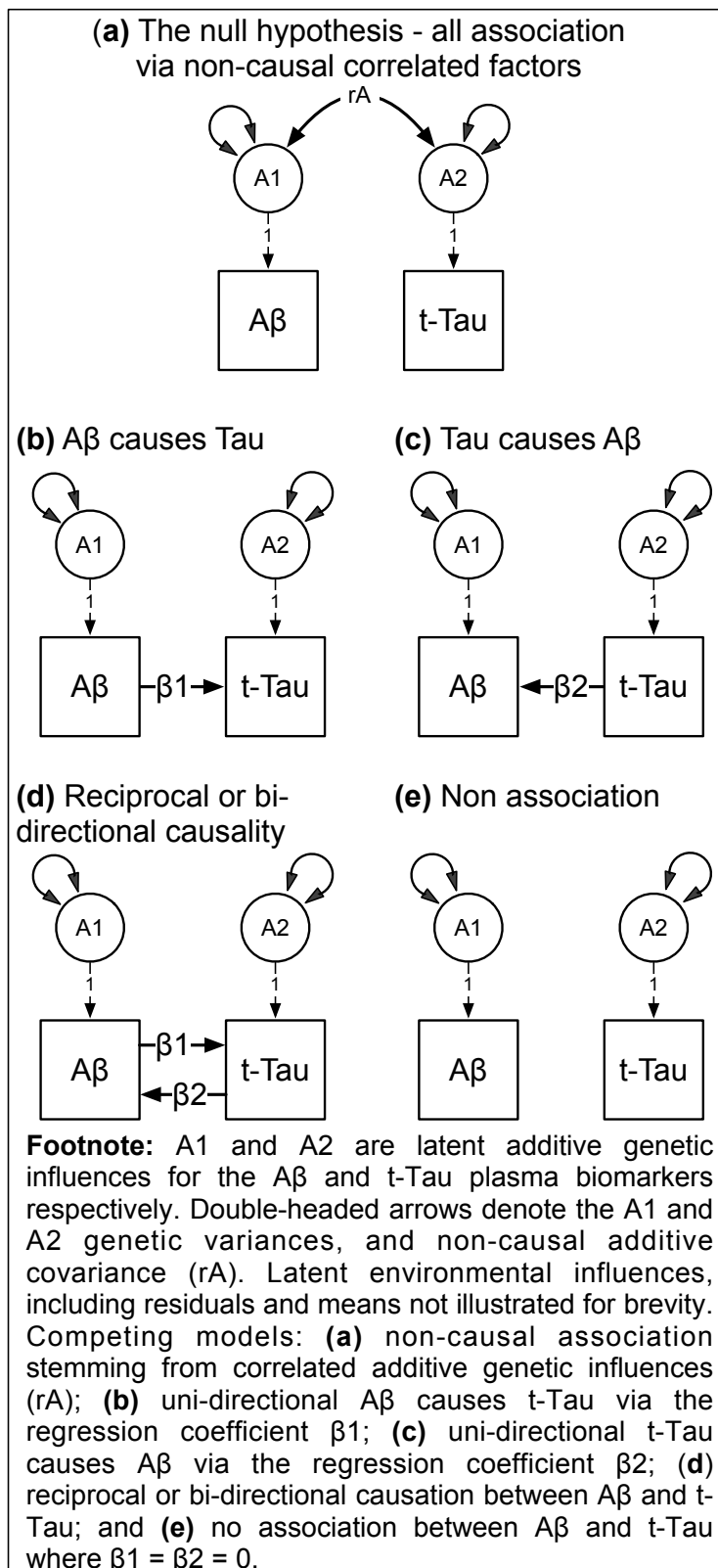
1. Hardy, J.A. and G.A. Higgins, *Alzheimer's disease: the amyloid cascade hypothesis*. *Science*, 1992. **256**(5054): p. 184-5.
2. Gillespie, N.A., et al., *The heritability of blood-based biomarkers related to risk of Alzheimer's Disease in a population-based sample of early old-age men*. *Alzheimer's & Dementia*, 2023. **20**(1): p. 356-65.
3. Duffy, D.L. and N.G. Martin, *Inferring the direction of causation in cross-sectional twin data: theoretical and empirical considerations [see comments]*. *Genetic Epidemiology*, 1994. **11**(6): p. 483-502.
4. Heath, A.C., et al., *Testing Hypotheses About Direction of Causation Using Cross-Sectional Family Data*. *Behavior Genetics*, 1993. **23**(1): p. 29-50.
5. Neale, M.C., et al., *Depression and parental bonding: cause, consequence, or genetic covariance?* *Genet Epidemiol*, 1994. **11**(6): p. 503-22.
6. Gillespie, N.A., et al., *Direction of causation modeling between cross-sectional measures of parenting and psychological distress in female twins*. *Behav Genet*, 2003. **33**(4): p. 383-96.
7. Gillespie, N.A., et al., *Modeling the direction of causation between cross-sectional measures of disrupted sleep, anxiety and depression in a sample of male and female Australian twins*. *J Sleep Res*, 2012. **21**(6): p. 675-83.
8. Bates, T.C., M. Neale, and H.H. Maes, *umx: A library for Structural Equation and Twin Modelling in R*. *Twin Research and Human Genetics*, 2019. **22**: p. 27-41.
9. R Development Core Team, *R: A language and environment for statistical computing*. *R Foundation for Statistical Computing, Vienna, Austria*. URL <https://www.R-project.org/>. 2018.
10. Neale, M.C. and L.R. Cardon, *Methodology for Genetic Studies of Twins and Families*. 1st ed. NATO ASI Series. 1992, Dordrecht: Kluwer Academic Publishers.
11. Boker, S., et al., *OpenMx: An Open Source Extended Structural Equation Modeling Framework*. *Psychometrika*, 2011. **76**(2): p. 306-317.
12. Akaike, H., *Factor analysis and AIC*. *Psychometrika*, 1987. **52**: p. 317-332.
13. Blennow, K. and H. Zetterberg, *Biomarkers for Alzheimer's disease: current status and prospects for the future*. *J Intern Med*, 2018. **284**(6): p. 643-663.

The genetics of blood-based AD biomarkers...

14. Blennow, K. and H. Zetterberg, *Fluid biomarker-based molecular phenotyping of Alzheimer's disease patients in research and clinical settings*. Prog Mol Biol Transl Sci, 2019. **168**: p. 3-23.
15. Coomans, E.M., et al., *Genetically identical twin-pair difference models support the amyloid cascade hypothesis*. Brain, 2023. **146**(9): p. 3735-3746.
16. Pérez-Grijalba, V., et al., *Plasma A β 42/40 ratio alone or combined with FDG-PET can accurately predict amyloid-PET positivity: a cross-sectional analysis from the AB255 Study*. Alzheimers Res Ther, 2019. **11**(1): p. 96.
17. Pérez-Grijalba, V., et al., *Plasma A β 42/40 Ratio Detects Early Stages of Alzheimer's Disease and Correlates with CSF and Neuroimaging Biomarkers in the AB255 Study*. J Prev Alzheimers Dis, 2019. **6**(1): p. 34-41.
18. Janelidze, S., et al., *Plasma β -amyloid in Alzheimer's disease and vascular disease*. Sci Rep, 2016. **6**: p. 26801.

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Figure 1. Competing hypothetical models to account for the association between the A β and t-Tau biomarkers.



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Table 1. Multivariate model fitting comparisons between the non-causal correlated factors reference model (a) and the two causal (b-c), bi-directional or reciprocal causation (d), and the no association (e) models.

	ep	-2LL	df	Δ -2LL	Δ df	p	AIC
(i) Aβ40 & Tau							
(a) Correlated / non-causal	11	4482.91	1930				4504.91
(b) A β 40 \rightarrow Tau	9	4484.34	1932	1.43	2	0.4895	4502.34
(c) Tau \rightarrow A β 40	9	4484.36	1932	1.45	2	0.4842	4502.36
(d) Reciprocal causation	10	4483.36	1931	0.45	1	0.5013	4503.36
(e) No association	8	4484.47	1933	1.55	3	0.6699	4500.47
(ii) Aβ42 & Tau							
(a) Correlated / non-causal	11	4423.18	1917				4445.18
(b) A β 42 \rightarrow Tau	9	4424.15	1919	0.98	2	0.6139	4442.15
(c) Tau \rightarrow A β 42	9	4423.65	1919	0.47	2	0.7891	4441.65
(d) Reciprocal causation	10	4423.35	1918	0.17	1	0.6775	4443.35
(e) No association	8	4439.72	1920	16.54	3	0.0009	4455.72
(iii) Both Aβs & Tau							
(a) Correlated / non-causal	21	6120.15	2910				6162.15
(b) A β s \rightarrow Tau	17	6123.54	2914	3.39	4	0.4946	6157.54
(c) Tau \rightarrow A β s	17	6122.68	2914	2.53	4	0.6390	6156.68
(d) Reciprocal causation	19	6120.86	2912	0.71	2	0.7003	6158.86
(e) No association	15	6157.59	2916	37.44	6	0.0000	6187.59

Note: ep = number of estimated parameters, -2LL = -2 x log-likelihood, Δ -2LL = change in -2 x log-likelihood, Δ df = change in degrees of freedom, AIC = Akaike Information Criteria. In each of the three analyses (i, ii & iii), the nested sub-models (b, c & d) each provided good fits to the data in terms of non-significant Δ -2LL & lower AIC values when compared to the null.