

FEATURED ARTICLE

# Optimal combinations of CSF biomarkers for predicting cognitive decline and clinical conversion in cognitively unimpaired participants and mild cognitive impairment patients: A multi-cohort study

Gemma Salvadó<sup>1</sup>  | Victoria Larsson<sup>1</sup> | Karly A Cody<sup>2</sup> | Nicholas C Cullen<sup>1</sup> | Erin M Jonaitis<sup>2,3</sup> | Erik Stomrud<sup>1,4</sup> | Gwendlyn Kollmorgen<sup>5</sup> | Norbert Wild<sup>5</sup> | Sebastian Palmqvist<sup>1,4</sup> | Shorena Janelidze<sup>1</sup> | Niklas Mattsson-Carlgren<sup>1,6,7</sup> | Henrik Zetterberg<sup>8,9,10,11,12</sup> | Kaj Blennow<sup>8,9</sup> | Sterling C Johnson<sup>2,3,13,14</sup> | Rik Ossenkoppele<sup>1,15</sup> | Oskar Hansson<sup>1,4</sup> 

<sup>1</sup>Clinical Memory Research Unit, Department of Clinical Sciences Malmö, Lund University, Lund, Sweden

<sup>2</sup>Wisconsin Alzheimer's Disease Research Center University of Wisconsin School of Medicine and Public Health Madison Wisconsin, University of Wisconsin-Madison, Madison, Wisconsin, USA

<sup>3</sup>Wisconsin Alzheimer's Institute, School of Medicine and Public Health, University of Wisconsin-Madison, Madison, Wisconsin, USA

<sup>4</sup>Memory Clinic, Skåne University Hospital, Malmö, Sweden

<sup>5</sup>Roche Diagnostics GmbH, Penzberg, Germany

<sup>6</sup>Wallenberg Center for Molecular Medicine, Lund University, Lund, Sweden

<sup>7</sup>Department of Neurology, Skåne University Hospital, Lund, Sweden

<sup>8</sup>Department of Psychiatry and Neurochemistry, the Sahlgrenska Academy at the University of Gothenburg, Mölndal, Sweden

<sup>9</sup>Clinical Neurochemistry Laboratory, Sahlgrenska University Hospital, Mölndal, Sweden

<sup>10</sup>Department of Neurodegenerative Disease, UCL Institute of Neurology, Queen Square, London, UK

<sup>11</sup>UK Dementia Research Institute at UCL, London, UK

<sup>12</sup>Hong Kong Center for Neurodegenerative Diseases, Hong Kong, China

<sup>13</sup>Department of Medicine, School of Medicine and Public Health, University of Wisconsin-Madison, Madison, Wisconsin, USA

<sup>14</sup>Geriatric Research, Education and Clinical Center at the William S. Middleton Memorial Veterans Hospital, University of Wisconsin-Madison, Madison, Wisconsin, USA

<sup>15</sup>Alzheimer Center Amsterdam, Department of Neurology, Amsterdam Neuroscience, Vrije Universiteit Amsterdam, Amsterdam University Medical Center, Amsterdam, The Netherlands

## Correspondence

Gemma Salvadó, Clinical Memory Research Unit, Department of Clinical Sciences SE-205 02, Malmö, Sweden.  
Email: [gemma.salvado@med.lu.se](mailto:gemma.salvado@med.lu.se)

Oskar Hansson, Memory Clinic, Skåne University Hospital, SE-205 02, Malmö, Sweden.  
Email: [oskar.hansson@med.lu.se](mailto:oskar.hansson@med.lu.se)

## Abstract

**Introduction:** Our objective was determining the optimal combinations of cerebrospinal fluid (CSF) biomarkers for predicting disease progression in Alzheimer's disease (AD) and other neurodegenerative diseases.

**Methods:** We included 1,983 participants from three different cohorts with longitudinal cognitive and clinical data, and baseline CSF levels of A $\beta$ 42, A $\beta$ 40, phosphorylated tau at threonine-181 (p-tau), neurofilament light (NfL), neurogranin,  $\alpha$ -synuclein,

Gemma Salvadó and Victoria Larsson are contributed equally as first authors.

This is an open access article under the terms of the [Creative Commons Attribution-NonCommercial](https://creativecommons.org/licenses/by-nc/4.0/) License, which permits use, distribution and reproduction in any medium, provided the original work is properly cited and is not used for commercial purposes.

© 2023 The Authors. *Alzheimer's & Dementia* published by Wiley Periodicals LLC on behalf of Alzheimer's Association.

**Funding information**

Swedish Research Council, Grant/Award Numbers: 2016-00906, 2018-02532, 2017-00915; Knut and Alice Wallenberg foundation, Grant/Award Number: 2017-0383; Marianne and Marcus Wallenberg foundation, Grant/Award Number: 2015.0125; Swedish Brain Foundation, Grant/Award Numbers: FO2021-0293, FO2020-0271; The Parkinson foundation of Sweden, Grant/Award Number: 1280/20; Skåne University Hospital Foundation, Grant/Award Number: 2020-O00028; Regionalt Forskningsstöd, Grant/Award Number: 2020-0314; Swedish federal government under the ALF agreement, Grant/Award Number: 2018-Projekt0279; Swedish Alzheimer Foundation, Grant/Award Numbers: AF-940046, AF-742881, AF-939932; European Research Council, Grant/Award Numbers: 681712, 101053962; Swedish State Support for Clinical Research, Grant/Award Number: ALFGBG-71320; Alzheimer Drug Discovery Foundation (ADDF), USA, Grant/Award Number: 201809-2016862; AD Strategic Fund and the Alzheimer's Association, Grant/Award Numbers: ADSF-21-831376-C, ADSF-21-831381-C, ADSF-21-831377-C; Olav Thon Foundation; Hjärnfonden, Sweden, Grant/Award Number: FO2019-0228; European Union's Horizon 2020 research and innovation programme under the Marie Skłodowska-Curie (MIRIADE), Grant/Award Number: 860197; European Union Joint Programme – Neurodegenerative Disease Research, Grant/Award Number: JPND2021-00694; UK Dementia Research Institute at UCL, Grant/Award Number: UKDRI-1003; Alzheimer Drug Discovery Foundation (ADDF), USA, Grant/Award Number: RDAPB-201809-2016615; Swedish government and the County Councils, Grant/Award Number: ALFGBG-715986; European Union Joint Program for Neurodegenerative Disorders, Grant/Award Number: JPND2019-466-236; NIH, Grant/Award Numbers: R01AG027161, R01AG021155, P30AG062715; University of Wisconsin Institute for Clinical and Translational Research NCATS, Grant/Award Number: TL1TR002375; the Strategic Research Area MultiPark (Multidisciplinary Research in Parkinson's disease) at Lund University; Erling-Persson Family Foundation; Stiftelsen för Gamla Tjänarinnor; Hjärnfonden, Sweden, Grant/Award Number: FO2017-0243

soluble triggering receptor expressed on myeloid cells 2 (sTREM2), glial fibrillary acidic protein (GFAP), YKL-40, S100b, and interleukin 6 (IL-6) (Elecys NeuroToolKit).

**Results:** Change of modified Preclinical Alzheimer's Cognitive Composite (mPACC) in cognitively unimpaired (CU) was best predicted by p-tau/A $\beta$ 42 alone ( $R^2 \geq 0.31$ ) or together with NfL ( $R^2 = 0.25$ ), while p-tau/A $\beta$ 42 ( $R^2 \geq 0.19$ ) was sufficient to accurately predict change of the Mini-Mental State Examination (MMSE) in mild cognitive impairment (MCI) patients. P-tau/A $\beta$ 42 (AUC  $\geq 0.87$ ) and p-tau/A $\beta$ 42 together with NfL (AUC  $\geq 0.75$ ) were the best predictors of conversion to AD and all-cause dementia, respectively.

**Discussion:** P-tau/A $\beta$ 42 is sufficient for predicting progression in AD, with very high accuracy. Adding NfL improves the prediction of all-cause dementia conversion and cognitive decline.

**KEYWORDS**

amyloid- $\beta$ , BioFINDER, cognitive decline, conversion to dementia, glial activation, inflammation, neurodegeneration, tau ratio, WADRC, WRAP

**1 | BACKGROUND**

Amyloid- $\beta$  (A $\beta$ ) plaques and neurofibrillary tau tangles are key pathological hallmarks in Alzheimer's disease (AD).<sup>1</sup> Cerebrospinal fluid (CSF) measures of these two pathologies have shown high accuracy for the diagnosis and prognosis of AD<sup>2,3</sup> and are currently, together with positron emission tomography (PET), used for diagnosis in the

clinical practice. CSF biomarker development has also enabled measurement of other pathophysiological alterations related to AD and other dementias. Some of these biomarkers may provide additional information on individuals' disease stage and progression, while others may enhance the understanding of underlying biological processes occurring during the course of the disease. Some of the most promising novel CSF biomarkers are related to neurodegeneration and microglial

**RESEARCH IN CONTEXT**

- 1. Systematic review:** There are currently many studies suggesting that biomarkers targeting pathophysiological alterations other than amyloid- $\beta$  and tau, like neurodegeneration and glial activation, may be associated with disease progression in Alzheimer's disease (AD) dementia and other neurodegenerative diseases. However, few have investigated the predictive power of the combination of these biomarkers with conventional measures of AD pathology.
- 2. Interpretation:** The best predictive accuracy for progression of AD and other neurodegenerative diseases using cerebrospinal fluid (CSF) biomarkers comes from using neurofilament light together with amyloid- $\beta$  and tau biomarkers.
- 3. Future directions:** Future studies need to investigate whether these results can be translated to a more diverse population as well as to plasma biomarkers.

or astroglial activation<sup>4</sup> (see<sup>5</sup> and<sup>6</sup> for extensive reviews). However, until recently, these biomarkers had to be measured in different platforms and using sophisticated methods, which reduced its applicability to clinical practice.

Some of these biomarkers have recently been implemented in a single panel of automated CSF immunoassays, with the Elecsys NeuroToolKit platform (Roche Diagnostics International Ltd, Rotkreuz, Switzerland), facilitating implementation and direct comparison between different centers. This panel measures biomarkers of neurodegeneration (neurofilament light [NfL], neurogranin and  $\alpha$ -synuclein); glial activity and neuroinflammation (soluble triggering receptor expressed on myeloid cells 2 [sTREM2], glial fibrillary acidic protein [GFAP], YKL-40 [also known as chitinase 3-protein 1], S100 calcium-binding protein B [S100b] and interleukin 6 [IL-6]); as well as core AD-biomarkers (phosphorylated tau at threonine-181 [p-tau], A $\beta$ 42, and A $\beta$ 40). Previous studies using this panel have shown alterations in some of these biomarkers in different stages and in relation to different aspects of the disease.<sup>7-10</sup> However, their clinical utility for predicting disease progression after accounting for core AD-biomarkers is still largely unknown. Studies in large cohorts from different centers are also needed to verify generalizability of findings related to these biomarkers.

Therefore, we aimed to identify the optimal combination of CSF biomarkers for an accurate prediction of cognitive decline and clinical conversion, using longitudinal data from three large cohorts. More specifically, we aimed to assess whether the core-AD biomarkers (operationalized as the CSF p-tau/A $\beta$ 42 ratio and the CSF A $\beta$ 42/40 ratio) would be sufficient to accurately predict disease progression or whether biomarkers targeting other pathophysiological mechanisms during AD could significantly improve their prediction.

**2 | METHODS****2.1 | Participants**

Participants were included from three cohorts: the Swedish BioFINDER-1 (NCT01208675)<sup>11</sup> and BioFINDER-2 (NCT03174938)<sup>12</sup> at Lund University (Lund, Sweden), and a joint cohort comprised of subsets who met inclusion criteria from Wisconsin Registry for Alzheimer's Prevention (WRAP) and Wisconsin Alzheimer's Disease Research Center (WADRC) at Wisconsin University (Wisconsin, USA).<sup>13</sup> In both BioFINDER-1 and BioFINDER-2 cognitively unimpaired (CU) people were recruited from population-based studies in the city of Malmö as previously described.<sup>11,12</sup> These individuals did not fulfil the criteria of mild cognitive impairment (MCI) or dementia at baseline. Furthermore, The BioFINDER studies included patients with either subjective cognitive decline (SCD) or MCI (none of one fulfilled the criteria for any type of dementia at baseline) from the Memory clinics in southern Sweden. Following the NIA-AA recommendations cognitive normal and subjects with SCD were defined as CU.<sup>14</sup> Exclusion criteria included (1) significant unstable systemic illness or organ failure, such as terminal cancer, that makes it difficult to participate in the study; (2) current significant alcohol or substance misuse; or (3) refusing lumbar puncture or neuropsychological assessment. Participants from WRAP and WADRC include CU and impaired participants, who are enriched for parental family history of AD, and undergo neuropsychological evaluations on an annual or biennial basis. More information about the recruitment of these participants can be found in,<sup>13</sup> but in summary, participants were included from memory clinics in which a parent was diagnosed or treated, advertisements, and word of mouth. The participants of the present study were enrolled between November 2007 to May 2020. All participants gave written informed consent and ethical approval was granted by the Regional Ethical Committee in Lund, Sweden, and University of Wisconsin Health Sciences Institutional Review Board, respectively.

**2.2 | CSF measurements**

CSF levels of A $\beta$ 42<sup>15</sup> as well as Total-Tau and Phospho-Tau(181P)<sup>16</sup> were measured using the Elecsys  $\beta$ -Amyloid (1-42), Total-Tau, and Phospho-Tau(181P) CSF electrochemiluminescence immunoassays on a fully automated cobas e 601 instrument (Roche Diagnostics International Ltd., Rotkreuz, Switzerland). The other CSF biomarkers (A $\beta$ 40, NfL, neurogranin, YKL-40, GFAP, sTREM2, S100b, IL-6, and  $\alpha$ -synuclein) were measured with robust prototype assays as part of the Roche NeuroToolKit on cobas e 411 and e 601 instruments (Roche Diagnostics International Ltd, Rotkreuz, Switzerland). All measurements were performed at the Clinical Neurochemistry Laboratory, University of Gothenburg, Sweden by board-certified laboratory technicians who were blinded to diagnostic and other clinical data. We also calculated the CSF p-tau/A $\beta$ 42 and the A $\beta$ 42/A $\beta$ 40 ratios. These ratios were included in this study as they may represent reliable

alternatives for amyloid PET<sup>17-21</sup> and are often used in the clinical setting for diagnostic<sup>22-26</sup> and/or prognostic purposes,<sup>27,28</sup> which is the main objective of this study. However, we acknowledge that these may not be a direct measure of AD-related brain pathology.

## 2.3 | Outcomes

The primary cognitive outcome for CU was the modified Preclinical Alzheimer's Cognitive Composite (mPACC) given its previously proven sensitivity to detect changes in cognition in CU participants.<sup>29</sup> The mPACC in BioFINDER-1 and BioFINDER-2 was calculated as the average of four z-scores, for tests of memory (the delayed recall test from the cognitive subscale from the Alzheimer's Disease Assessment Scale [ADAS-cog]) with a double weight to preserve the weight of memory tests in the original PACC,<sup>29</sup> verbal ability (animal fluency), executive function (Trail Making Test B [TMT-B]), and global cognition using the Mini-Mental Estate Examination (MMSE), as previously described.<sup>30</sup> In WRAP and WADRC, mPACC was calculated as the mean of z-scores of three tests: TMT-B for timed executive function, delayed story recall harmonized from either the Craft Story (for WADRC) or WMS-R Logical Memory story A (for WRAP), and the total over trials of the Rey Auditory-Verbal Learning Test (AVLT).<sup>31</sup> For MCI patients, cognitive decline was measured with MMSE in all cohorts. In the joint Wisconsin cohort some MMSE scores were obtained through conversion from Montreal Cognitive Assessment (MOCA) scores as previously described.<sup>32</sup>

## 2.4 | Statistical analyses

Two main analyses were performed aiming at identifying the optimal combination of CSF biomarkers to predict either cognitive decline or clinical conversion. The biomarkers included as individual predictors in all the analyses were: the p-tau/A $\beta$ 42 ratio, the A $\beta$ 42/40 ratio, NfL, neurogranin, YKL-40, GFAP, sTREM2, S100b, IL-6, and  $\alpha$ -synuclein. To study cognitive decline, we used mPACC for CU individuals and MMSE for MCI participants, independently for each cohort. Linear mixed models were used to predict cognitive decline using random intercept and random time slopes with baseline age, sex, APOE- $\epsilon$ 4 carriership, and education as covariates. The model including only covariates and time was considered the basic model. To select the best model including baseline CSF biomarkers, a forward selection procedure was used based on Bayesian information criteria (BIC) using the *aba* package for R (v.4.1.0). For all biomarkers included in the model, we also included their interaction with time. For each case (i.e., cognitive measure and cohort), a parsimonious model (i.e., best prediction with the lowest number of predictors) was selected with a forward step-wise inclusion of the most significant biomarkers until the difference in BIC with the next model (i.e., including an additional biomarker) was not significant ( $\Delta$ BIC > 6), which indicates strong evidence of improvement.<sup>33</sup> When different, the parsimonious models were compared with those only

including covariates and CSF p-tau/A $\beta$ 42 ratio, as a typical measure of AD pathology used in the clinical setting.<sup>2</sup>

For predicting clinical conversion, we followed a similar approach using generalized linear models including a binomial family. Progression to AD dementia and to all-cause dementia were studied independently in each cohort for both CU and MCI. Models with less than 20 cases progressing to dementia were disregarded. Basic models included age, sex, and APOE- $\epsilon$ 4 carriership as covariates. Best models were selected also using a forward selection procedure with the *aba* package based on differences in area under the curve (AUC). Models were selected as preferred based on AUC using the DeLong's test to check for significant differences ( $p < 0.05$ ). Best models were also compared with those only including covariates and CSF p-tau/A $\beta$ 42 ratio. We also computed hazard ratios (HRs) for each biomarker selected in each of the parsimonious models using cox proportional hazards regression models, and constructed Kaplan-Meier curves using the *survival* package from R. Finally, we repeated all the main analyses with data truncated at 4- and 6-years of follow-up.

## 3 | RESULTS

### 3.1 | Participants

A total of 1,453 CU and 530 MCI patients were included in this study. The mean (SD) age of CU individuals was 66.6 (9.7) years, 857 (59.0%) of them were women, and 553 (38.1%) were APOE- $\epsilon$ 4 carriers. For MCI patients, the mean (SD) age was 71.2 (7.2) years, and there were 225 (42.5%) women and 269 (50.8%) APOE- $\epsilon$ 4 carriers. CU were followed for a mean (SD) time of 4.7 (3.1) years and MCI for 2.8 (2.2) years. Baseline characteristics are presented in Table 1. Out of the information available, 43/1000 (4.3%) CU participants converted to AD dementia and 66/1000 (6.6%) to all-cause dementia. From MCI at baseline patients, 185/529 (35.0%) converted to Alzheimer's dementia and 297/529 (56.1%) to all-cause dementia. Cross-correlation of CSF biomarkers by cohort can be found in Figure S1.

### 3.2 | Cognitive decline

Parsimonious models for predicting cognitive decline can be found in Table 2. In summary, mPACC change was best predicted by CSF p-tau/A $\beta$ 42 ratio alone in BioFINDER-2 and WRAP & WADRC (BioFINDER-2:  $\beta$ [95%CI] = -0.10 [-0.14, -0.05],  $R^2 = 0.36$ ; WRAP & WADRC:  $\beta$ [95%CI] = -0.12 [-0.15, -0.09],  $R^2 = 0.31$ ). In BioFINDER-1, the addition of CSF NfL ( $\beta$ [95%CI] = -0.09 [-0.12, -0.05]) to CSF p-tau/A $\beta$ 42 ratio ( $\beta$ [95%CI] = -0.15 [-0.19, -0.12]) significantly improved this prediction (with NfL:  $R^2 = 0.25$  vs. without NfL:  $R^2 = 0.20$ ,  $\Delta$ BIC = 26,  $p < 0.001$ ). The parsimonious model to predict change in MMSE in MCI patients only included CSF p-tau/A $\beta$ 42 ratio in all three cohorts (BioFINDER-1:  $\beta$ [95%CI] = -0.24 [-0.30, -0.17],  $R^2 = 0.40$ ; BioFINDER-2:  $\beta$ [95%CI] = -0.14 [-0.23, -0.05]  $R^2 = 0.19$ ; WRAP &

**TABLE 1** Participant characteristics

|   | BioFINDER-1                 |                             | BioFINDER-2                 |                             | WRAP and WADRC              |                             |
|---|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|
|   | CU (n = 550)                | MCI (n = 260)               | CU (n = 453)                | MCI (n = 215)               | CU (n = 450)                | MCI (n = 55)                |
| Age, mean (SD)<br>[range]                                 | 71.8 (6.09)<br>[41.3, 88.4] | 71.5 (5.44)<br>[60.1, 80.8] | 65.7 (11.6)<br>[40.5, 88.7] | 70.9 (8.37)<br>[43.1, 93.3] | 61.2 (7.64)<br>[42.7, 85.5] | 70.6 (9.07)<br>[54.1, 87.3] |
| Women, No (%)   | 312 (56.7%)                 | 107 (41.2%)                 | 242 (53.4%)                 | 98 (45.6%)                  | 303 (67.3%)                 | 20 (36.4%)                  |
| APOE-ε4 carriers,<br>No (%)                               | 181 (32.9%)                 | 129 (49.6%)                 | 204 (45.0%)                 | 113 (52.6%)                 | 168 (37.3%)                 | 27 (49.1%)                  |
| Years of<br>education,<br>mean (SD)                       | 12.4 (3.6)                  | 11.0 (3.3)                  | 12.6 (3.4)                  | 12.5 (4.1)                  | 16.2 (2.5)                  | 15.9 (2.7)                  |
| Years of<br>follow-up, mean<br>(SD)                       | 6.2 (2.2) [0,<br>10.9]      | 3.3 (2.1) [0.35,<br>8.5]    | 1.2 (1.04) [0,<br>3.6]      | 1.5 (1.10) [0,<br>3.6]      | 6.6 (2.34) [0.9,<br>11.2]   | 5.2 (2.47) [1.1,<br>10.0]   |
| Cognitive<br>measure, mean<br>(SD)*                       | 0.041 (0.719)               | 27.0 (1.9)                  | 0.028 (0.764)               | 27.0 (1.9)                  | 0.108 (0.704)               | 27.6 (2.1)                  |
| Missing, No (%)   | 96 (17.5%)                  | -                           | 5 (1.1%)                    | 1 (0.5%)                    | 33 (7.3%)                   | 1 (1.8%)                    |
| Conversion to<br>Alzheimer's-<br>type dementia,<br>No (%) | 39 (7.1%)                   | 117 (45.0%)                 | -                           | 50 (23.3%)                  | 4 (0.9%)                    | 18 (32.7%)                  |
| Missing, No (%)   | -                           | 1 (0.4%)                    | -                           | -                           | -                           | -                           |
| Conversion to<br>all-cause<br>dementia, No<br>(%)         | 61 (11.1%)                  | 189 (72.7%)                 | -                           | 84 (39.1%)                  | 5 (1.1%)                    | 24 (43.6%)                  |
| Missing, No (%)   | -                           | 1 (0.4%)                    | -                           | -                           | -                           | -                           |

\*Cognitive measure was mPACC for CU and MMSE for MCI patients.

Abbreviations: CU, cognitively unimpaired; MCI, mild cognitive impairment; MMSE, Mini-Mental State Examination; mPACC, modified Preclinical Alzheimer Cognitive Composite; WADRC, Wisconsin Alzheimer's Disease Research Center; WRAP, Wisconsin Registry for Alzheimer's Prevention.

**TABLE 2** Parsimonious model description for predicting cognitive decline

| Diagnosis group | Cohort       | Biomarker          | $\beta$ [95% CI]     | BIC parsimonious | R <sup>2</sup> parsimonious | BIC basic | R <sup>2</sup> basic |
|-----------------|--------------|--------------------|----------------------|------------------|-----------------------------|-----------|----------------------|
| CU              | BioFINDER-1  | p-tau/A $\beta$ 42 | -0.15 [-0.19, -0.12] | 3109.2           | 0.25                        | 3230.6    | 0.06                 |
|                 |              | NfL                | -0.09 [-0.12, -0.05] |                  |                             |           |                      |
|                 | BioFINDER-2  | p-tau/A $\beta$ 42 | -0.10 [-0.14, -0.05] | 1297.9           | 0.36                        |           |                      |
| MCI             | WRAP & WADRC | p-tau/A $\beta$ 42 | -0.12 [-0.15, -0.09] | 3631.6           | 0.31                        | 3681.8    | 0.26                 |
|                 | BioFINDER-1  | p-tau/A $\beta$ 42 | -0.24 [-0.30, -0.17] | 2390.7           | 0.40                        | 2426.9    | 0.30                 |
|                 | BioFINDER-2  | p-tau/A $\beta$ 42 | -0.14 [-0.23, -0.05] | 976.6            | 0.19                        | 978.8     | 0.13                 |
|                 | WRAP & WADRC | p-tau/A $\beta$ 42 | -0.32 [-0.49, -0.15] | 582.0            | 0.32                        | 594.79    | 0.14                 |

Description of the parsimonious model for predicting cognitive decline for each diagnosis group and cohort. These models were derived independently per each cohort and baseline diagnosis. Only those biomarkers selected to be into the parsimonious model per cohort and disease stage are shown in each case. Effect sizes ( $\beta$ ) for each selected biomarker and statistics (BIC and R<sup>2</sup>) of the final model are shown. BIC and R<sup>2</sup> of the basic model (only covariates) are included for comparison. Linear mixed models with random slope and intercept were used in all cases.  $\beta$  estimates represent the effect size of each biomarker's interaction with time. Cognitive decline was assessed with mPACC in CU and with MMSE in MCI patients. Covariates were included in all models and were: age, sex, APOE-ε4 carriership, education and time.

Abbreviations: A $\beta$ , amyloid- $\beta$ ; BIC, Bayesian Information Criterion; CI, confidence interval; CU, cognitively unimpaired; MCI, mild cognitive impairment; MMSE, Mini-Mental State Examination; mPACC, modified Preclinical Alzheimer Cognitive Composite; p-tau, phosphorylated tau; WADRC, Wisconsin Alzheimer's Disease Research Center; WRAP, Wisconsin Registry for Alzheimer's Prevention.

WADRC:  $\beta$ [95%CI] = -0.32 [-0.49, -0.15],  $R^2 = 0.32$ ). However, in BioFINDER-2 this model was not significantly different than the basic model only including covariates. Depiction of cognitive decline and predictions using the parsimonious models are shown in Figure 1. Sensitivity analyses with cognitive data truncated at 4 and 6 years of follow-up can be found in Tables S1 and S2, respectively.

### 3.3 | Clinical conversion

Parsimonious models for predicting conversion to AD and all-cause dementia are summarized in Table 3. Conversion to AD dementia was best predicted by CSF p-tau/A $\beta$ 42 ratio both in CU participants (BioFINDER-1 CU: AUC[95%CI] = 0.95 [0.93 - 0.97]) and MCI patients at baseline in the cohorts with data available (BioFINDER-1 MCI: AUC[95%CI] = 0.92 [0.89 - 0.95]; BioFINDER-2 MCI: AUC[95%CI] = 0.87[0.82 - 0.92]).

In the case of conversion to all-cause dementia, CSF NfL and CSF p-tau/A $\beta$ 42 ratio were included in the parsimonious model in CU participants at baseline (BioFINDER-1 CU: AUC[95%CI] = 0.90[0.86 - 0.95]). In participants that were MCI patients at baseline, in two out of three cohorts the CSF p-tau/A $\beta$ 42 ratio was included in the parsimonious model (BioFINDER-1 and WRAP & WADRC). In BioFINDER-1, CSF NfL was also included in the parsimonious model (BioFINDER-1 MCI: AUC[95%CI] = 0.83 [0.77 - 0.89]). In WRAP & WADRC, the CSF p-tau/A $\beta$ 42 ratio alone was sufficient to accurately predict conversion to all-cause dementia in MCI patients at baseline (AUC[95%CI] = 0.83[0.72 - 0.94]). In BioFINDER-2, the parsimonious model to predict conversion from MCI to all-cause dementia included the A $\beta$ 42/40 ratio and CSF NfL (AUC[95%CI] = 0.75[0.68 - 0.82]). However, this model was not significantly better than the one including the p-tau/A $\beta$ 42 ratio and NfL (AUC[95%CI] = 0.74[0.67 - 0.81],  $p = 0.383$ ).

We additionally calculated the HR for a one-SD increase in all CSF biomarkers selected in each of the parsimonious (Table 3). In summary, the CSF p-tau/A $\beta$ 42 ratio HRs [95% CI] for predicting conversion to AD dementia were 2.12 [1.80 - 2.49] in CU participants and around 2 in MCI patients at baseline (BioFINDER-1: HR[95%CI] = 1.94 [1.68 - 2.25]; BioFINDER-2: HR[95%CI] = 2.35 [1.84 - 3.08]). In the case of predicting conversion to all-cause dementia, HRs for the CSF p-tau/A $\beta$ 42 ratio were all around 1.5 both for CU participants (BioFINDER-1 CU: 1.65 [1.44 - 1.90]) and MCI patients at baseline (BioFINDER-1 MCI: 1.45 [1.23 - 1.63]; WRAP & WADRC MCI: 1.69[1.05 - 2.72]). HRs for NfL were all above 1.4 also in CU participants (BioFINDER-1 CU: 2.09 [1.68 - 2.61]) and MCI patients at baseline (BioFINDER-1 MCI: 1.42[1.23 - 1.63]; BioFINDER-2 MCI: 1.52[1.25 - 1.84]). Conversely, HR for the A $\beta$ 42/40 ratio (BioFINDER-2: 0.58[0.46 - 0.72]) was lower than 1 when predicting all-cause dementia from MCI. This was due to the well-known a negative association between the levels of this biomarker and pathology.

Receiver operating characteristic (ROC) curves for parsimonious, CSF p-tau/A $\beta$ 42 ratio-only, and basic (i.e., only covariates) models are shown in Figure 2 and Kaplan-Meier curves for all the parsimonious

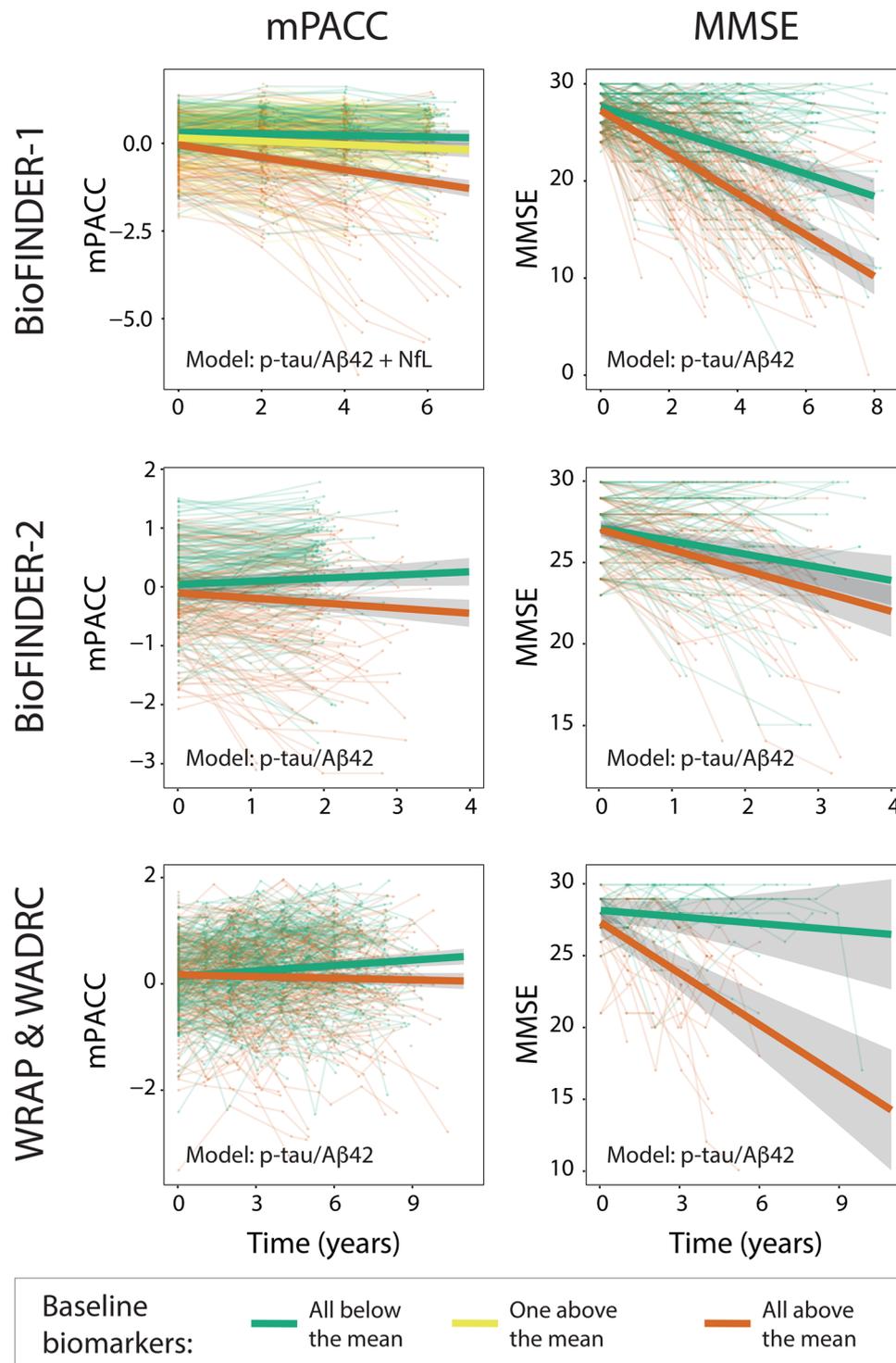
models in Figure 3. Sensitivity analyses with conversion data truncated at 4- and 6-years follow-up can be found in Tables S3 and S4, respectively. Sensitivity analyses where the p-tau/A $\beta$ 42 ratio was excluded are shown in Table S5. All AUCs were equivalent or lower than models with the p-tau/A $\beta$ 42 ratio. To visualize this, we also included comparison plots between the individual markers and the p-tau/A $\beta$ 42 and A $\beta$ 42/40 ratios for BioFINDER-1 participants based on whether or not they converted to Alzheimer's dementia (Figure S2).

## 4 | DISCUSSION

In this study, we have sought to identify the optimal combination of several CSF biomarkers analyzed with recently developed fully automated assays for predicting disease progression in three large longitudinal cohorts. We found that the CSF p-tau/A $\beta$ 42 ratio alone or in combination with CSF NfL, a marker of neurodegeneration, may be sufficient to accurately predict disease progression both in CU participants and in MCI patients. The addition of CSF NfL may be especially important to predict conversion to all-cause dementia and, in some cases, cognitive decline. In contrast, other neurodegenerative markers and glial-related biomarkers did not significantly improve our predictive models. Furthermore, we showed that the CSF p-tau/A $\beta$ 42 ratio was preferentially selected in the parsimonious models over the CSF A $\beta$ 42/40 ratio as a prognosis biomarker in AD. Although many previous studies have shown individual associations between some of these biomarkers and progression in AD and other dementias, this is the first study to investigate their prognosis utility as a combination of biomarkers and to compare it to typical core-AD biomarkers. Notably, our results have been replicated in three large independent cohorts with relatively long follow-up. Altogether, our study supports the use of the CSF p-tau/A $\beta$ 42 ratio, together with CSF NfL, in the clinical setting for prognosis of AD and other dementias.

The finding of CSF p-tau/A $\beta$ 42 ratio as the best marker for predicting disease progression was not unexpected. For many years now, the CSF p-tau/A $\beta$ 42 ratio has already shown high accuracy in predicting disease progression in both CU participants<sup>34,35</sup> and MCI patients.<sup>2,36</sup> The novelty of our study is that among a large panel of established and more novel CSF biomarkers, the CSF p-tau/A $\beta$ 42 ratio alone may be sufficient for an accurate disease prognosis, except for the addition of CSF NfL in some cases. Furthermore, we also showed that the CSF p-tau/A $\beta$ 42 ratio may be a more useful metric than the CSF A $\beta$ 42/40 ratio as a prognosis tool in AD, although the latter might be conceptually more sound as the former is combining two non-linearly associated processes.<sup>21</sup> These results have important ramifications for clinical settings, where these biomarkers are increasingly available.

As aforementioned, we also found that CSF NfL was included in the parsimonious models for predicting disease progression in particular scenarios. In this study, higher levels of CSF NfL have consistently reported higher hazard ratios for converting to *all-cause* dementia, both for CU and MCI groups. In line with our results, previous studies have already shown the tight association between CSF NfL and brain and cognitive deterioration.<sup>37</sup> The fact that CSF NfL improved the



**FIGURE 1** Depiction of cognitive change over time per cohort and clinical group. Cognitive change in CU participants is shown with mPACC change (first column) while cognitive change in MCI patients is shown with MMSE (second column). Bold lines represent the predicted trajectory using the parsimonious model based on biomarkers levels at baseline. Green (red) lines represent participants with all biomarkers in the parsimonious model below (above) the mean at baseline. For the model with more than one biomarker included in the parsimonious model, the yellow line represents subjects with only one biomarker above the mean at baseline. Gray bands represent 95% confidence intervals. These lines are only for visualization purposes and represent the mean trajectory of the group of participants included in these artificial groups. Individual trajectories were calculated per each participant in the statistical model. Linear mixed models with random slope and intercept were used to construct the models. Biomarkers used in the parsimonious model in each case are detailed in the plots. Abbreviations: CU, cognitively unimpaired; MCI, mild cognitive impairment; MMSE, Mini-Mental State Examination; mPACC, modified Preclinical Alzheimer Cognitive Composite; WADRC, Wisconsin Alzheimer's disease Research Center; WRAP, Wisconsin Registry for Alzheimer's Prevention

**TABLE 3** Parsimonious model description for predicting clinical conversion

| Conversion type             | Cohort       | Biomarker  | HR [95%CI]        | p-value | AUC [95%CI]<br>Parsimonious<br>model | AUC [95%CI]<br>basic model |
|-----------------------------|--------------|------------|-------------------|---------|--------------------------------------|----------------------------|
| CU → AD<br>dementia         | BioFINDER-1  | p-tau/aβ42 | 2.12 [1.80, 2.49] | <0.001  | 0.95 [0.93-0.97]                     | 0.78 [0.70-0.86]           |
| MCI → AD<br>dementia        | BioFINDER-1  | p-tau/aβ42 | 1.94 [1.68, 2.25] | <0.001  | 0.92 [0.89-0.95]                     | 0.77 [0.71-0.83]           |
|                             | BioFINDER-2  | p-tau/aβ42 | 2.35 [1.80, 3.08] | <0.001  | 0.87 [0.82-0.92]                     | 0.75 [0.68-0.82]           |
| CU → All-cause<br>dementia  | BioFINDER-1  | p-tau/aβ42 | 1.65 [1.44, 1.90] | <0.001  | 0.90 [0.86-0.95]                     | 0.74 [0.67-0.80]           |
|                             |              | NfL        | 2.09 [1.68, 2.61] | <0.001  |                                      |                            |
| MCI → All-cause<br>dementia | BioFINDER-1  | p-tau/aβ42 | 1.45 [1.26, 1.67] | <0.001  | 0.83 [0.77-0.89]                     | 0.71 [0.63-0.78]           |
|                             |              | NfL        | 1.42 [1.23, 1.63] | <0.001  |                                      |                            |
|                             | BioFINDER-2  | aβ42/40    | 0.58 [0.42, 0.79] | <0.001  | 0.75 [0.68-0.82]                     | 0.62 [0.54-0.70]           |
|                             |              | NfL        | 1.52 [1.25, 1.84] | <0.001  |                                      |                            |
|                             | WRAP & WADRC | p-tau/aβ42 | 1.69 [1.05, 2.72] | 0.032   | 0.83 [0.72-0.94]                     | 0.73 [0.59-0.87]           |

Description of the parsimonious model for predicting clinical conversion for each conversion type and cohort. These models were derived independently per each cohort and baseline diagnosis. Only those biomarkers selected to be into the parsimonious model per cohort and disease stage are shown in each case. Hazard ratios (HRs) for each selected biomarker and AUC of the final model are shown. AUCs of the basic model (only covariates) are included for comparison. HRs were calculated with cox proportional hazards regression model with clinical conversion as the outcome. For all biomarkers, HRs represent increased risk of conversion for each SD change in biomarker value. To calculate AUC, we used generalized linear models including a binomial family. Covariates included in all models were: age, sex, and APOE-ε4 carriership.

Abbreviations: AD, Alzheimer's disease; AUC, area under the curve; Aβ, amyloid-β; CI, confidence interval; CU, cognitively unimpaired; HR, hazard ratio; MCI, mild cognitive impairment; NfL, neurofilament light; phosphorylated tau; p-tau; SD, standard deviation; WADRC, Wisconsin Alzheimer's Disease Research Center; WRAP, Wisconsin Registry for Alzheimer's Prevention.

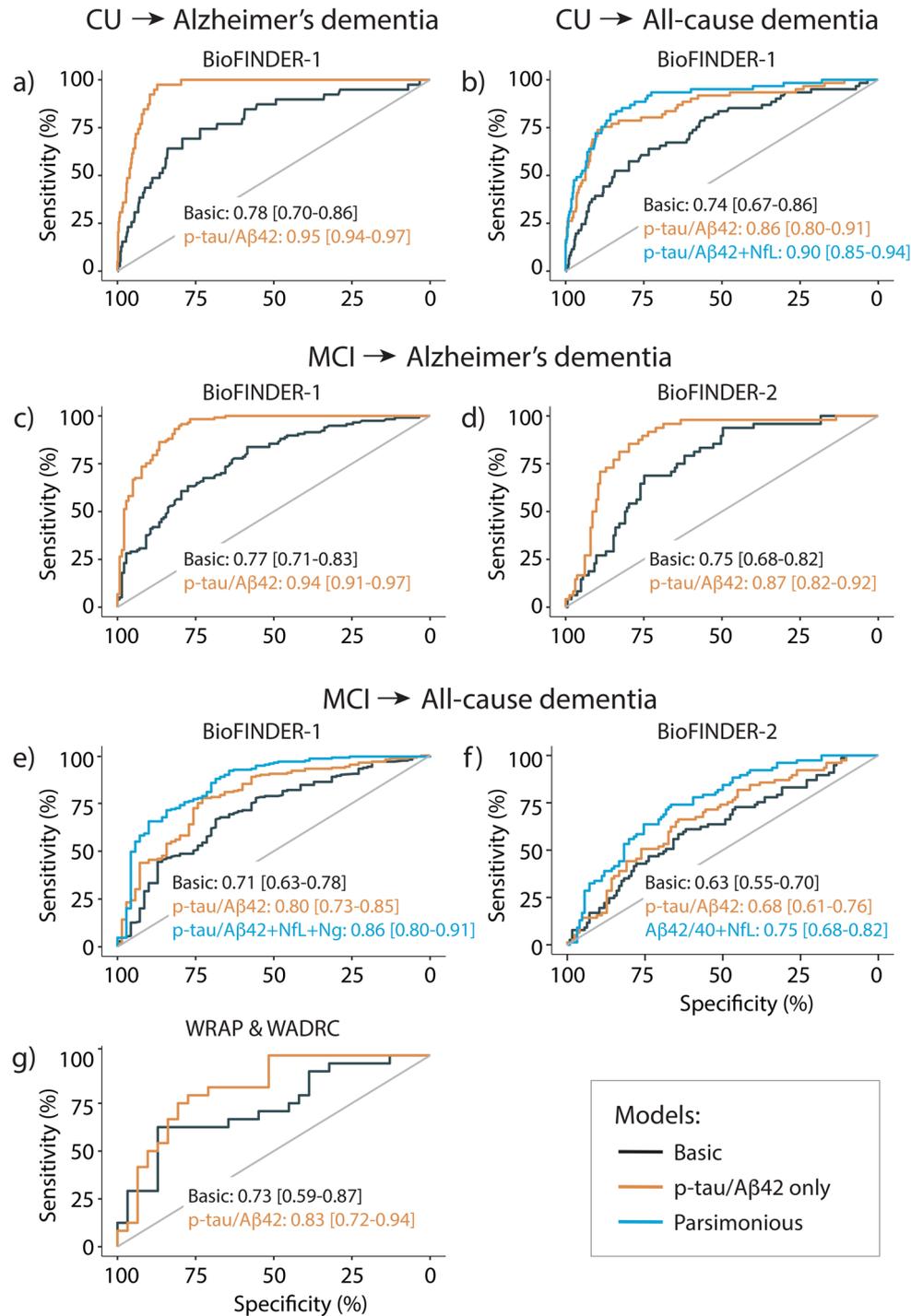
prediction models of conversion to *all-cause* dementia but not to *Alzheimer's* dementia was in line with previous findings of CSF NfL as a non-specific biomarker of neurodegeneration.<sup>5,38</sup> Thus, based on our results, CSF NfL may be helpful to predict conversion to dementias other than Alzheimer's, which in our models was already sufficiently captured by the CSF p-tau/Aβ42 ratio.

We found that the addition of CSF NfL to the CSF p-tau/Aβ42 ratio, was also useful for predicting cognitive decline. This was only significant in the case of mPACC change prediction in BioFINDER-1 ( $n = 550$ ). However, this CU group had a longer follow-up than in BioFINDER-2 (BioFINDER-1 mean follow-up time: 6.2 years vs. BioFINDER-2 mean follow-up time: 1.2 years) and was significantly older than in the joint WRAP & WADRC group (BioFINDER-1 mean age: 71.8 years old vs. WRAP & WADRC mean age: 61.2 years old). These characteristics may have facilitated the detection of a larger change in cognition, which was not only attributed to CSF p-tau/Aβ42 ratio. Further supporting the utility of CSF NfL for predicting cognitive decline, it is important to note that it was the most frequently selected biomarker, after the CSF p-tau/Aβ42 ratio, although in most cases its addition did not significantly improve the model. Nonetheless, we cannot disregard the possibility of a false positive result for CSF NfL. Future studies may benefit from including a longer follow-up and a more diverse population to elucidate the full clinical value of CSF NfL.

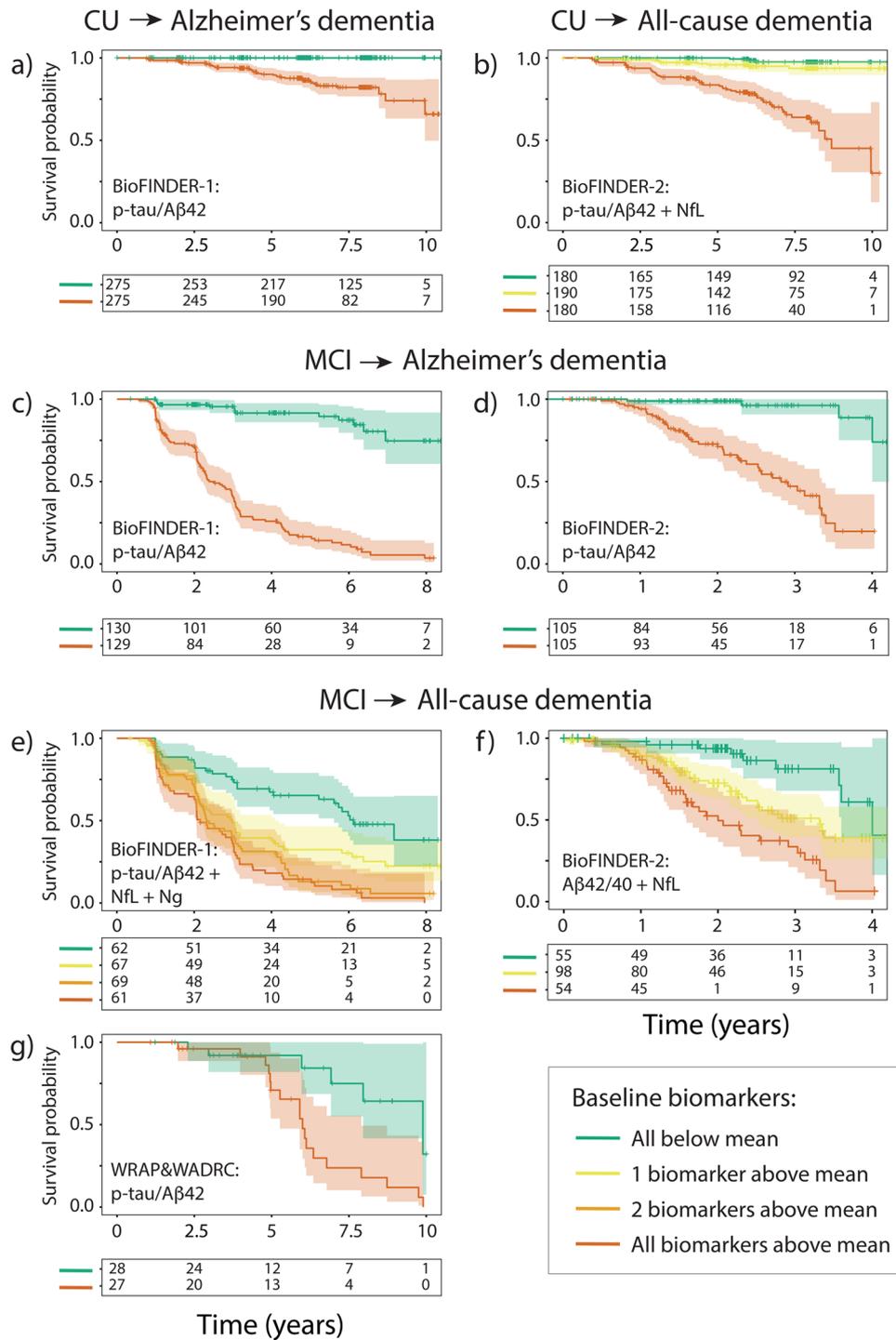
In the recent years, there has been an increasing interest in investigating glial activation markers and their relationship with disease progression, as it has been suggested that inflammatory processes may have an active role in AD.<sup>4</sup> Previous studies proposed that some glial activation markers could be related to disease progression.<sup>7,8</sup> For

instance, higher levels of sTREM2, a microglial activation marker,<sup>39</sup> have been related to slower rates of Aβ accumulation and disease progression.<sup>40-42</sup> Levels of YKL-40, an astrocytic marker, were also proposed as a potential prognostic marker of AD.<sup>43</sup> However, those studies were conducted without their comparison to core AD biomarkers or if so, they did not show a significant improvement on their prediction accuracy. Similarly, neurodegenerative markers other than CSF NfL were not included in our final models. Altogether, our results suggest that, although their study hold great importance to understand the biological processes underlying the course of AD, glial activation markers and neurodegeneration markers other than NfL do not provide additional value in predicting disease progression to justify their use in a clinical setting.

The main strength of this study is that analyses were performed in three large longitudinal cohorts. Although there were small differences in particular models, parsimonious models for each condition were replicated in all cohorts, reinforcing our results. Furthermore, we used a set of CSF biomarkers measured with the same technique and in the same single panel, which reduces variability in our measures. Nonetheless, some limitations must be acknowledged. First, we focused on CSF biomarkers rather than the recently developed plasma biomarkers. Plasma biomarkers are promising but are, in contrast with CSF, not yet widely available in clinical practice.<sup>12,44-47</sup> Second, participants were recruited from cohort studies, which may limit the generalizability of our results to a more representative clinical population. Nonetheless, BioFINDER-2 is a study not only focused on AD and include participants with all types of dementias; and BioFINDER-1 is a representative sample of patients from a Memory Clinic in Sweden, which increases



**FIGURE 2** Depiction of ROC curves for basic, p-tau/aβ42 only and parsimonious models for predicting progression to AD or all-cause dementia. Models were created independently per clinical group at baseline (CU or MCI) and cohort (BioFINDER-1: [a], [c], and [e]; BioFINDER-2: [b], [d] and [f]; and WRAP & WADRC: [g]). Only scenarios with data available and more than 20 conversion cases are depicted. AUC[95%CI] are depicted for each model and case in the picture. Basic models only included covariates. Biomarkers included in the parsimonious models in each case are detailed in the figure. Covariates were included in all models and were: age, sex, APOE-ε4 carriership and time. Abbreviations: Aβ, amyloid-β; AD, Alzheimer's disease; AUC, area under the curve; CI, confidence interval; CU, cognitively unimpaired; MCI, mild cognitive impairment; NfL, neurofilament light; Ng, neurogranin; p-tau, phosphorylated tau; ROC, receiver operating characteristic; WADRC, Wisconsin Alzheimer's disease Research Center; WRAP, Wisconsin Registry for Alzheimer's Prevention



**FIGURE 3** Kaplan-Meier survival curves for each cohort and clinical group at baseline with conversion to Alzheimer's dementia or all-cause dementia as outcome. Colored lines depict different groups of individuals based on their baseline levels of selected biomarkers in the parsimonious models. Green (red) group line represents participants with all biomarkers in the parsimonious model below (above) the mean at baseline. For models with more than one biomarker included in the parsimonious model, yellow lines represent subjects with only one biomarkers above the mean at baseline. Colored bands represent 95% confidence intervals and crosses represent censored data. The total number of individuals in each group and timepoints are shown in the tables below the curves. Biomarkers used in the parsimonious model in each case are detailed in the plots. Abbreviations: Aβ, amyloid-β; CU, cognitively unimpaired; MCI, mild cognitive impairment; NfL, neurofilament light; Ng, neurogranin; p-tau; phosphorylated tau; WADRC, Wisconsin Alzheimer's disease Research Center; WRAP, Wisconsin Registry for Alzheimer's Prevention

the diversity of the sample. Third, although our mean of follow-up time was considerably long, it may still not be sufficient to capture cognitive decline in CU, as most neurodegenerative disorders show a relatively gradual decline. To alleviate this caveat, we used the mPACC to measure cognitive decline in CU, which is specifically developed to detect the first signs of cognitive decline in a cognitively normal population.<sup>29</sup> Nonetheless, results of clinical progression in this group may be biased due to the low number of conversions, specially to Alzheimer's dementia, and for this reason our results should be considered with caution. Finally, we acknowledge that although the p-tau/A $\beta$ 42 ratio showed prognostic utility, it may not be the most adequate biomarker to assess the actual brain pathology.

Our results suggest that the CSF p-tau/A $\beta$ 42 ratio is sufficient, compared with other CSF biomarkers, to accurately predict disease progression in three large longitudinal cohorts. The addition of CSF NfL to the CSF p-tau/A $\beta$ 42 ratio may improve prediction of progression to all-cause dementia and cognitive decline in some cases. On the other hand, other markers of neurodegeneration and glial activation markers do not seem to provide additional value on disease progression prediction. These results may be useful for its implementation to the clinical setting although further research in more diverse populations is needed.

## ACKNOWLEDGMENTS

Work at the authors' research center was supported by the Swedish Research Council (2016-00906), the Knut and Alice Wallenberg foundation (2017-0383), the Marianne and Marcus Wallenberg foundation (2015.0125), the Strategic Research Area MultiPark (Multidisciplinary Research in Parkinson's disease) at Lund University, the Swedish Alzheimer Foundation (AF-939932), the Swedish Brain Foundation (FO2021-0293), The Parkinson foundation of Sweden (1280/20), the Konung Gustaf V:s och Drottning Victorias Frimurare-stiftelse, the Skåne University Hospital Foundation (2020-000028), Regionalt Forskningsstöd (2020-0314) and the Swedish federal government under the ALF agreement (2018-Projekt0279). SP is supported by Swedish Research Council (2018-02052), the Swedish Brain Foundation (FO2020-0271) and the Swedish Alzheimer Foundation (AF-940046). HZ is a Wallenberg Scholar supported by grants from the Swedish Research Council (#2018-02532), the European Research Council (#681712 and #101053962), Swedish State Support for Clinical Research (#ALFGBG-71320), the Alzheimer Drug Discovery Foundation (ADDF), USA (#201809-2016862), the AD Strategic Fund and the Alzheimer's Association (#ADSF-21-831376-C, #ADSF-21-831381-C and #ADSF-21-831377-C), the Olav Thon Foundation, the Erling-Persson Family Foundation, Stiftelsen för Gamla Tjänarinnor, Hjärnfonden, Sweden (#FO2019-0228), the European Union's Horizon 2020 research and innovation programme under the Marie Skłodowska-Curie grant agreement No 860197 (MIRIADE), the European Union Joint Programme - Neurodegenerative Disease Research (JPND2021-00694), and the UK Dementia Research Institute at UCL (UKDRI-1003). KB is supported by the Swedish Research Council (#2017-00915), the Alzheimer Drug Discovery Foundation (ADDF), USA (#RDAPB-201809-2016615), the

Swedish Alzheimer Foundation (#AF-742881), Hjärnfonden, Sweden (#FO2017-0243), the Swedish state under the agreement between the Swedish government and the County Councils, the ALF-agreement (#ALFGBG-715986), and European Union Joint Program for Neurodegenerative Disorders (JPND2019-466-236). The University of Wisconsin authors (KAC, EMJ, SCJ) and data were supported by NIH (R01AG027161, R01AG021155, P30AG062715), and the University of Wisconsin Institute for Clinical and Translational Research NCATS (TL1TR002375).

The funding sources had no role in the design and conduct of the study; in the collection, analysis, interpretation of the data; or in the preparation, review, or approval of the manuscript.

## CONFLICT OF INTEREST

O.H. has acquired research support (for the institution) from ADx, AVID Radiopharmaceuticals, Biogen, Eli Lilly, Eisai, Fujirebio, GE Healthcare, Pfizer, and Roche. In the past 2 years, he has received consultancy/speaker fees from Amylyx, Alzpath, BioArctic, Biogen, Cerveau, Fujirebio, Genentech, Novartis, Roche, and Siemens. G.K. and N.W. are employees of Roche Diagnostics. SP has served on scientific advisory boards and/or given lectures in symposia sponsored by Biogen, Eli Lilly, Geras Solutions, and Roche. H.Z. has served at scientific advisory boards and/or as a consultant for Abbvie, Alektor, Annexon, Artery Therapeutics, AZTherapies, CogRx, Denali, Eisai, Nervgen, Novo Nordisk, Pinteon Therapeutics, Red Abbey Labs, Passage Bio, Roche, Samumed, Siemens Healthineers, Triplet Therapeutics, and Wave, has given lectures in symposia sponsored by Cellectricon, Fujirebio, Alzecure, Biogen, and Roche, and is a co-founder of Brain Biomarker Solutions in Gothenburg AB (BBS), which is a part of the GU Ventures Incubator Program (outside submitted work). K.B. has served as a consultant, at advisory boards, or at data monitoring committees for Abcam, Axon, Biogen, JOMDD/Shimadzu, Julius Clinical, Lilly, MagQu, Novartis, Prothena, Roche Diagnostics, and Siemens Healthineers, and is a co-founder of Brain Biomarker Solutions in Gothenburg AB (BBS), which is a part of the GU Ventures Incubator Program. S.C.J. previously served on an advisory board for Roche Diagnostics, and receives research funding from NIH and from Cerveau Technologies. COBAS, COBAS E and ELECSYS are trademarks of Roche. The Elecsys  $\beta$ -Amyloid (1-42) CSF assay, the Elecsys Phospho-Tau (181P) CSF assay, and the Elecsys Total-Tau CSF assay are not approved for clinical use in the United States. The NeuroToolKit robust prototype assays are for investigational purposes and are not approved for clinical use. G.S.B., V.L., K.A.C., N.C.C., E.M.J., E.S., S.J., N.M.C., and R.O. have nothing to disclose. Author disclosures are available in the [supporting information](#).

## ORCID

Gemma Salvadó  <https://orcid.org/0000-0002-5210-9230>

Oskar Hansson  <https://orcid.org/0000-0001-8467-7286>

## REFERENCES

1. Jack CR, Bennett DA, Blennow K, et al. NIA-AA Research Framework: toward a biological definition of Alzheimer's disease. *Alzheimer's & Dementia*. 2018;14:535-562. doi:10.1016/j.jalz.2018.02.018

2. Hansson O, Zetterberg H, Buchhave P, Londos E, Blennow K, Minthon L. Association between CSF biomarkers and incipient Alzheimer's disease in patients with mild cognitive impairment: a follow-up study. *Lancet Neurol*. 2006;5:228-234. doi:10.1016/S1474-4422(06)70355-6
3. Blennow K, Shaw LM, Stomrud E, et al. Predicting clinical decline and conversion to Alzheimer's disease or dementia using novel Elecsys A $\beta$ (1-42), pTau and tTau CSF immunoassays. *Sci Rep*. 2019;9:1-11. doi:10.1038/s41598-019-54204-z
4. Heneka MT, Carson MJ, Khoury J El, et al. Neuroinflammation in Alzheimer's disease. *Lancet Neurol*. 2015;14:388-405. doi:10.1016/S1474-4422(15)70016-5
5. Milà-Alomà M, Suárez-Calvet M, Molinuevo JL. Latest advances in cerebrospinal fluid and blood biomarkers of Alzheimer's disease. *Ther Adv Neurol Disord*. 2019;12. doi:10.1177/1756286419888819
6. Hansson O. Biomarkers for neurodegenerative diseases. *Nat Med*. 2021;27:954-963. doi:10.1038/s41591-021-01382-x
7. Salvadó G, Milà-Alomà M, Shekari M, et al. Cerebral amyloid- $\beta$  load is associated with neurodegeneration and gliosis: mediation by p-tau and interactions with risk factors early in the Alzheimer's continuum. *Alzheimer's and Dementia*. 2021;1-13. doi:10.1002/alz.12245
8. Milà-Alomà M, Salvadó G, Gispert JD, et al. Amyloid- $\beta$ , tau, synaptic, neurodegeneration and glial biomarkers in the preclinical stage of the Alzheimer's continuum. *Alzheimer's and Dementia*. 2020;1-14. doi:10.1002/alz.12131
9. van Hulle C, Jonaitis EM, Betthausen TJ, et al. An examination of a novel multipanel of CSF biomarkers in the Alzheimer's disease clinical and pathological continuum. *Alzheimer's and Dementia*. 2021;17:431-445. doi:10.1002/alz.12204
10. Salvadó G, Shekari M, Falcon C, et al. Brain alterations in the early Alzheimer's continuum with amyloid- $\beta$ , tau, glial and neurodegeneration CSF markers. *Brain Commun*. 2022;4. doi:10.1093/braincomms/fcac134
11. Palmqvist S, Tideman P, Cullen N, et al. Prediction of future Alzheimer's disease dementia using plasma phospho-tau combined with other accessible measures. *Nat Med*. 2021. doi:10.1038/s41591-021-01348-z
12. Palmqvist S, Janelidze S, Quiroz YT, et al. Discriminative accuracy of plasma phospho-tau217 for Alzheimer disease vs other neurodegenerative disorders. *JAMA - Journal of the American Medical Association*. 2020;324:772-781. doi:10.1001/jama.2020.12134
13. Johnson SC, Kosciak RL, Jonaitis EM, et al. The Wisconsin registry for Alzheimer's prevention: a review of findings and current directions. *Alzheimers Dement (Amst)*. 2018;10:130-142. doi:10.1016/j.dadm.2017.11.007
14. Petersen RC, Wiste HJ, Weigand SD, et al. NIA-AA Alzheimer's Disease Framework: clinical characterization of stages. *Ann Neurol*. 2021;89:1145-1156. doi:10.1002/ana.26071
15. Bittner T, Zetterberg H, Teunissen CE, et al. Technical performance of a novel, fully automated electrochemiluminescence immunoassay for the quantitation of  $\beta$ -amyloid (1-42) in human cerebrospinal fluid. *Alzheimer's and Dementia*. 2016;12:517-526. doi:10.1016/j.jalz.2015.09.009
16. Lifke V, Kollmorgen G, Manuilova E, et al. Elecsys® Total-Tau and Phospho-Tau (181P) CSF assays: analytical performance of the novel, fully automated immunoassays for quantification of tau proteins in human cerebrospinal fluid. *Clin Biochem*. 2019;72:30-38. doi:10.1016/j.clinbiochem.2019.05.005
17. Keshavan A, Wellington H, Chen Z, et al. Concordance of CSF measures of Alzheimer's pathology with amyloid PET status in a preclinical cohort: a comparison of Lumipulse and established immunoassays. *Alzheimers Dement (Amst)*. 2021;13:e12131. doi:10.1002/dad2.12131
18. Kaplow J, Vandijck M, Gray J, et al. Concordance of Lumipulse cerebrospinal fluid t-tau/A $\beta$ 42 ratio with amyloid PET status. *Alzheimers Dement*. 2020;16:144-152. doi:10.1002/alz.12000
19. Janelidze S, Pannee J, Mikulskis A, et al. Concordance between different amyloid immunoassays and visual amyloid positron emission tomographic assessment. *JAMA Neurol*. 2017;74:1492-1501. doi:10.1001/jamaneurol.2017.2814
20. Alcolea D, Pegueroles J, Muñoz L, et al. Agreement of amyloid PET and CSF biomarkers for Alzheimer's disease on Lumipulse. *Ann Clin Transl Neurol*. 2019;6:1815-1824. doi:10.1002/acn3.50873
21. Hansson O, Lehmann S, Otto M, Zetterberg H, Lewczuk P. Advantages and disadvantages of the use of the CSF Amyloid  $\beta$  (A $\beta$ ) 42/40 ratio in the diagnosis of Alzheimer's Disease. *Alzheimers Res Ther*. 2019;11:1-15. doi:10.1186/s13195-019-0485-0
22. Leitão MJ, Silva-Spínola A, Santana I, et al. Clinical validation of the Lumipulse G cerebrospinal fluid assays for routine diagnosis of Alzheimer's disease. *Alzheimers Res Ther*. 2019;11:91. doi:10.1186/s13195-019-0550-8
23. Bayart J-L, Hanseeuw B, Ivanoiu A, van Pesch V. Analytical and clinical performances of the automated Lumipulse cerebrospinal fluid A $\beta$ 42 and T-Tau assays for Alzheimer's disease diagnosis. *J Neurol*. 2019;266:2304-2311. doi:10.1007/s00415-019-09418-6
24. Paciotti S, Sepe FN, Eusebi P, et al. Diagnostic performance of a fully automated chemiluminescent enzyme immunoassay for Alzheimer's disease diagnosis. *Clin Chim Acta*. 2019;494:74-78. doi:10.1016/j.cca.2019.03.1612
25. Paterson RW, Slattery CF, Poole T, et al. Cerebrospinal fluid in the differential diagnosis of Alzheimer's disease: clinical utility of an extended panel of biomarkers in a specialist cognitive clinic. *Alzheimers Res Ther*. 2018;10:32. doi:10.1186/s13195-018-0361-3
26. Mattsson-Carlgen N, Grinberg LT, Boxer A, et al. Cerebrospinal fluid biomarkers in autopsy-confirmed Alzheimer disease and frontotemporal lobar degeneration. *Neurology*. 2022;98:e1137-50. doi:10.1212/WNL.0000000000200040
27. Blennow K, Shaw LM, Stomrud E, et al. Predicting clinical decline and conversion to Alzheimer's disease or dementia using novel Elecsys A $\beta$ (1-42), pTau and tTau CSF immunoassays. *Sci Rep*. 2019;9:19024. doi:10.1038/s41598-019-54204-z
28. Hansson O, Seibyl J, Stomrud E, et al. CSF biomarkers of Alzheimer's disease concord with amyloid- $\beta$  PET and predict clinical progression: a study of fully automated immunoassays in BioFINDER and ADNI cohorts. *Alzheimer's and Dementia*. 2018;1-12. doi:10.1016/j.jalz.2018.01.010
29. Donohue MC, Sperling RA, Salmon DP, et al. The preclinical Alzheimer cognitive composite: measuring Amyloid-Related Decline. *JAMA Neurol*. 2014;71:961-970. doi:10.1001/jamaneurol.2014.803
30. Mattsson-Carlgen N, Janelidze S, Palmqvist S, et al. Longitudinal plasma p-tau217 is increased in early stages of Alzheimer's disease. *Brain*. 2020;143:3234-3241. doi:10.1093/BRAIN/AWAA286
31. Jonaitis EM, Kosciak RL, Clark LR, et al. Measuring longitudinal cognition: individual tests versus composites. *Alzheimer's & Dementia: Diagnosis, Assessment & Disease Monitoring*. 2019;11:74-84. doi:10.1016/j.dadm.2018.11.006
32. Monsell SE, Dodge HH, Zhou X-H, et al. Results from the NACC uniform data set neuropsychological battery crosswalk study. *Alzheimer Dis Assoc Disord*;30:134-139. doi:10.1097/WAD.000000000000111.n.d.
33. Raftery AE. Bayesian model selection in social research. *Sociol Methodol*. 1995;25:111-163. doi:10.2307/271063
34. Li G, Sokal I, Quinn JF, et al. CSF tau/A 42 ratio for increased risk of mild cognitive impairment: a follow-up study. *Neurology*. 2007;69:631-639. doi:10.1212/01.wnl.0000267428.62582.aa
35. Fagan AM, Roe CM, Xiong C, Mintun MA, Morris JC, Holtzman DM. Cerebrospinal fluid tau/ $\beta$ -Amyloid42 ratio as a prediction of cognitive decline in nondemented older adults. *Arch Neurol*. 2007;64:343. doi:10.1001/archneur.64.3.noc60123
36. Snider BJ, Fagan AM, Roe C, et al. Cerebrospinal fluid biomarkers and rate of cognitive decline in very mild dementia of the

- Alzheimer type. *Arch Neurol.* 2009;66. doi:[10.1001/archneurol.2009.55](https://doi.org/10.1001/archneurol.2009.55)
37. Zetterberg H, Skillbäck T, Mattsson N, et al. Association of cerebrospinal fluid neurofilament light concentration with Alzheimer disease progression. *JAMA Neurol.* 2016;73:60-67. doi:[10.1001/jamaneurol.2015.3037](https://doi.org/10.1001/jamaneurol.2015.3037)
  38. Bridel C, van Wieringen WN, Zetterberg H, et al. Diagnostic value of cerebrospinal fluid neurofilament light protein in neurology: a systematic review and meta-analysis. *JAMA Neurol.* 2019;76:1035-1048. doi:[10.1001/jamaneurol.2019.1534](https://doi.org/10.1001/jamaneurol.2019.1534)
  39. Suárez-Calvet M, Kleinberger G, Araque Caballero MÁ, et al. sTREM2 cerebrospinal fluid levels are a potential biomarker for microglia activity in early-stage Alzheimer's disease and associate with neuronal injury markers. *EMBO Mol Med.* 2016;8:466-476. doi:[10.15252/emmm.201506123](https://doi.org/10.15252/emmm.201506123)
  40. Ewers M, Franzmeier N, Suárez-Calvet M, et al. Increased soluble TREM2 in cerebrospinal fluid is associated with reduced cognitive and clinical decline in Alzheimer's disease. *Sci Transl Med.* 2019;11. doi:[10.1126/scitranslmed.aav6221](https://doi.org/10.1126/scitranslmed.aav6221)
  41. Ewers M, Biechele G, Suárez-Calvet M, et al. Higher CSF sTREM2 and microglia activation are associated with slower rates of beta-amyloid accumulation. *EMBO Mol Med.* 2020;12:e12308. doi:[10.15252/emmm.202012308](https://doi.org/10.15252/emmm.202012308)
  42. Morenas-Rodríguez E, Li Y, Nuscher B, et al. Soluble TREM2 in CSF and its association with other biomarkers and cognition in autosomal-dominant Alzheimer's disease: a longitudinal observational study. *Lancet Neurol.* 2022;21:329-341. doi:[10.1016/S1474-4422\(22\)00027-8](https://doi.org/10.1016/S1474-4422(22)00027-8)
  43. Craig-Schapiro R, Perrin RJ, Roe CM, et al. YKL-40: a novel prognostic fluid biomarker for preclinical Alzheimer's disease. *Biol Psychiatry.* 2010;68:903-912. doi:[10.1016/j.biopsych.2010.08.025](https://doi.org/10.1016/j.biopsych.2010.08.025)
  44. Benedet AL, Milà-Alomà M, Vrillon A, et al. Differences between plasma and cerebrospinal fluid glial fibrillary acidic protein levels across the Alzheimer disease continuum. *JAMA Neurol.* 2021;78:1471-1483. doi:[10.1001/jamaneurol.2021.3671](https://doi.org/10.1001/jamaneurol.2021.3671)
  45. Thijssen EH, la Joie R, Wolf A, et al. Diagnostic value of plasma phosphorylated tau181 in Alzheimer's disease and frontotemporal lobar degeneration. *Nat Med.* 2020;26:387-397. doi:[10.1038/s41591-020-0762-2](https://doi.org/10.1038/s41591-020-0762-2)
  46. Karikari TK, Pascoal TA, Ashton NJ, et al. Blood phosphorylated tau 181 as a biomarker for Alzheimer's disease: a diagnostic performance and prediction modelling study using data from four prospective cohorts. *Lancet Neurol.* 2020;19:422-433. doi:[10.1016/S1474-4422\(20\)30071-5](https://doi.org/10.1016/S1474-4422(20)30071-5)
  47. Verberk IMW, Thijssen E, Koelewijn J, et al. Combination of plasma amyloid beta(1-42/1-40) and glial fibrillary acidic protein strongly associates with cerebral amyloid pathology. *Alzheimers Res Ther.* 2020;12:118. doi:[10.1186/s13195-020-00682-7](https://doi.org/10.1186/s13195-020-00682-7)

## SUPPORTING INFORMATION

Additional supporting information can be found online in the Supporting Information section at the end of this article.

**How to cite this article:** Salvadó G, Larsson V, Cody KA, et al. Optimal combinations of CSF biomarkers for predicting cognitive decline and clinical conversion in cognitively unimpaired participants and mild cognitive impairment patients: A multi-cohort study. *Alzheimer's Dement.* 2023;1-13. <https://doi.org/10.1002/alz.12907>