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Blood protein predictors of brain amyloid for enrichment in clinical trials?

Nicholas J. Ashton^{1, 2#}, Steven J. Kiddle^{1, 3#}, John Graf⁴, Malcolm Ward⁵, Alison Baird^{1, 6}, Abdul Hye^{1, 2}, Sarah Westwood^{1, 2}, Karyuan Vivian Wong¹, Richard J. Dobson^{1, 2}, Gil D. Rabinovici⁷, Bruce L. Miller⁷, Howard J. Rosen⁷, Andrew Torres⁴, Zhanpan Zhang⁴, Lennart Thurfjell⁸, Antonia Covin⁹, Cristina Tan Hehir⁴, David Baker⁹, Chantal Bazenet^{1, 2} and Simon Lovestone^{1, 6} and the AIBL Research Group¹⁰.

¹Institute of Psychiatry, King's College London, London, UK

²NIHR Biomedical Research Centre for Mental Health and Biomedical Research Unit for Dementia at South London and Maudsley NHS Foundation, London, UK

³MRC Social, Genetic and Developmental Psychiatry Centre, King's College London, London, UK

⁴GE Global Research, Niskayuna, NY

⁵Proteomics Facility, Institute of Psychiatry, Kings College London

⁶Department of Psychiatry, University of Oxford, Oxford, UK.

⁷Memory & Aging Center, Dept. of Neurology, University of California San Francisco⁷ GE Healthcare Life Sciences, Uppsala, Sweden.

⁸GE Healthcare Life Sciences, Uppsala, Sweden.

⁹Janssen R&D, Neurosciences, Titusville, NJ

¹⁰www.AIBL.csiro.au.

these authors contributed equally to this manuscript

Abstract

Background: Measures of neocortical amyloid burden (NAB) identify individuals who are at substantially greater risk for developing Alzheimer's disease (AD). Blood-based biomarkers predicting NAB would have great utility for enrichment of AD clinical trials, including large-scale prevention trials.

Methods: Non-targeted proteomic discovery was applied to 78 subjects from Australian Imaging, Biomarkers and Lifestyle Flagship Study of Ageing (AIBL) with a range of NAB values. Technical and independent replications were performed by immunoassay.

Results: Seventeen discovery candidates were selected for technical replication. Alpha-2-macroglobulin (α -2m), fibrinogen gamma chain (FGG) and complement factor H-related protein 1 (FHR-1) were confirmed to be associated with NAB. In an independent cohort FGG plasma levels combined with age predicted NAB with a sensitivity of 59% and specificity of 78%.

Conclusions: A single blood protein – FGG – combined with age was shown to relate to NAB and therefore could have potential for enrichment of clinical trials populations.

Keywords: Plasma, β amyloid, proteomics, Alzheimer's disease, biomarker, fibrinogen gamma chain and clinical trials.

Background

The diagnosis of AD can only be confirmed, with certainty, by histological examination of brain tissue at autopsy. This inspection should demonstrate considerable evidence of the classical pathological hallmarks of AD; extracellular amyloid beta ($A\beta$) plaques and intracellular neurofibrillary tangles predominantly comprised of hyperphosphorylated Tau [1]. Although an age-related disease usually affecting people over the age of 65 it is believed the accumulation of $A\beta$ plaques begins 15-20 years prior to clinical presentation [2] and reaches a plateau when cognitive, functional and behavioral decline occurs [3]. Existing treatments for AD are only capable of temporary symptomatic relief in a subset of patients [4]. As elevated brain $A\beta$ is an important risk factor for eventual AD, it has become critical to identify individuals at the early stages of $A\beta$ deposition to recruit into clinical trials of potentially disease-modifying therapeutics. Indeed, three prevention trials in asymptomatic individuals at the early stages of $A\beta$ deposition have commenced recently [5].

At present neuroimaging and cerebrospinal fluid (CSF) biomarkers are the accepted standards used to provide evidence of on-going AD pathophysiology related to $A\beta$ plaques. [^{11}C]-Pittsburgh Compound B (PiB) coupled with Positron Emission Tomography (PET) is widely used in research in measuring *in vivo* $A\beta$ deposition as its uptake in AD correlates with $A\beta$ plaques measured neuropathologically in the same brains [6]. The availability of longer lived [^{18}F]-labeled $A\beta$ PET tracers, such as flutemetamol [7] and florbetapir [8] could foster wider utilization in clinical use [9]. Early “proof of concept” PiB-PET studies demonstrated an increase of $A\beta$ deposition in a majority of individuals clinically diagnosed with AD as judged by visual assessment [10] or quantification of tracer uptake [11,12]. Two large studies, from Victoria

(Australia) and the University of California San Francisco Memory and Aging Center (UCSF, USA) have shown that PiB PET could discriminate between AD and non-A β dementias [11,13]. Some, but not all [14,15], studies also show that amyloid deposition as measured using PiB-PET either predicts decline in cognitive measures or tracks with such [2,16].

Many disease modifying therapeutics being developed target amyloid generation, deposition or clearance [17]. Recent phase III trials targeting amyloid reported that approximately 20% of trial participants actually had little or no A β when studied later using such PET imaging (Suspected Non Amyloid Pathology; SNAP) [18]. This is a very serious problem for such trials – success is hard to find in the field of neurodegeneration but likely to be significantly harder when a large minority of trial subjects fail to have the primary target pathology.

A solution is to use amyloid-PET scans (~\$3,000 per scan) to ensure primary target pathology. The first study to use this will be the Anti Amyloid in Asymptomatic AD (A4; n=1,000) prevention trial. In A4 the screen failure rate is anticipated to be even higher (~66%) due to the use of asymptomatic subjects. The great expense of the anticipated ~20% and ~66% amyloid-PET screen failure rates for clinical and prevention anti-amyloid trials, means that a blood test with even relatively low predictive accuracy for NAB has the potential to greatly reduce costs. This would work by applying the blood tests to large numbers of potentially eligible subjects, and only performing PET scans on those whose blood tests are positive. This would reduce the screen failure rates, and save money if the blood test was inexpensive comparatively. Therefore, a blood-based measure that correlates with neocortical amyloid burden (NAB) would be of considerable value as an enrichment filter for clinical trials.

The obvious blood candidate biomarker of brain A β pathology would be A β itself. A systematic literature review and meta-analysis by Koyama et al [19], on 10,303 subjects, found that lower plasma A β 42:A β 40 ratios were significantly associated with development of AD. However, the estimates had wide confidence intervals, due to high inter-study differences. As such plasma A β 42:A β 40 ratios are unlikely to be useful by itself for the prediction of NAB. The same study found that individual A β 42 and A β 40 levels in blood were not significantly associated with AD. Clearly novel biomarkers are needed that reflect brain amyloid pathology in blood.

There has been considerable effort in the search for AD blood-based biomarkers. Most studies use a case-control design, based on a clinical diagnosis of AD as determined by medical history, cognitive assessments and clinical examination. This classical, case versus age-matched controls approach, has identified a large number of putative plasma biomarkers (reviewed in [20,21,22]). However, such approaches are intrinsically flawed in the context of AD where a considerable proportion of cognitively unimpaired controls will be in the prodromal phase of AD, e.g. asymptomatic but with elevated NAB.

An approach to overcome this is to use a non-apparent measure of disease activity (endophenotype paradigm). The endophenotype approach is increasingly being adopted, for example to study blood-based biomarkers of cognitive decline [23,24], APOE4 risk [25], brain atrophy [26,27] and hippocampal metabolism [28]. More recently, blood-based biomarkers of NAB, as measured by PiB PET, have been reported [29,30,31]. Both Kiddle et al and Burnham et al utilized the Rules Based Medicine panel of 190 analytes to discover plasma proteins that related to NAB, and proposed a thirteen and five analyte model respectively. These models both contained the protein pancreatic polypeptide.

In a different approach Thambisetty et al used two dimensional gel electrophoresis (2D-GE) coupled with mass spectrometry (MS) to identify protein spots associating with NAB in an unbiased fashion. This study identified 6 proteins for spots associated with NAB, including APOE and Complement C3 which were independently replicated in the Kiddle et al study. 2D-GE is a well established technique for blood biomarker research and offers many advantages. However, it is restricted by a lengthy procedure with poor reproducibility that can only indentify a small number of “candidate spots” in limited sample sets.

In this study we employ a methodology that combines the unbiased approach of gel-based proteomics with high-throughput multiplex technology and the latest in MS instrumentation. This has enabled the identification and quantification of several hundred proteins, comparable to some panel based arrays, without losing the key advantages of unbiased gel-based discovery. This is the first application of this approach to identify blood-based biomarkers of NAB, and was applied to a subset of patients from the AIBL cohort with either high or low NAB. Promising markers were then replicated using immunoassays, first in the same cohort and then in an independent cohort [13].

Material and Methods

The Australian Imaging, Biomarkers and Lifestyle Flagship Study of Ageing (AIBL)

The AIBL study is a longitudinal study of ageing, neuroimaging, biomarkers, lifestyle, clinical and neuropsychological analysis with a focus on early detection and lifestyle intervention

(<http://www.aibl.csiro.au/>). Additional specifics regarding subject recruitment, diagnosis and study design have been described previously [32].

Discovery cohort; assessments, blood collection and processing

In total we examined plasma samples from a subset of 78 subjects from the AIBL study, who had undergone PiB-PET scans. A standardized uptake value ratio (SUVR) cutoff of 1.3 was used to classify subjects as belonging to PiB+ and PiB- groups. To increase statistical power the subjects were selected to be enriched for clear cases of PiB negativity and positivity. Standardised clinical assessments included Mini-Mental State Examination (MMSE), and Apolipoprotein E (APOE) genotypes were available.

The details of blood collection and sample processing have been previously discussed [31]. Plasma proteomic analysis and immunoassay measures were undertaken at King's College London (KCL).

AIBL PiB-PET

The PiB imaging methodology of the AIBL study is detailed elsewhere [33]. SUVR were generated using the cerebellar grey matter as the reference region as described in Burnham et al [31]. NAB was expressed as the average SUVR of the mean of frontal, superior parietal, lateral temporal, lateral occipital, and anterior and posterior cingulate regions.

Tandem Mass Tag (TMT) protein labeling, enzymatic digestion and peptide extraction

Each sample was randomly assigned and labeled with an Amine-Reactive TMT Reagent (TMT127-TMT131; Thermo Scientific #90064) with TMT126 being used to label the study

reference, an equal pool of the plasma obtained from all 78 subjects. A complete TMT6Plex combined five labeled plasma samples with a labeled study reference. In general, sample preparation and TMT labeling was performed as previously described [23,34] with some minor modifications (Supplementary Methods 1). Each TMT6Plex underwent 1D gel electrophoresis and excised into 10 fractions (Supplementary Methods 2). Gel pieces were then de-stained, digested, peptides extracted and lyophilised to completion prior to MS analysis (Supplementary Methods 3). LC-MS/MS data separated with 1D-GE can show a single protein in multiple fractions. Therefore, identical protein identifications observed in different fractions were considered as separate entities defined as protein molecular weight (MW) isoforms (Supplementary Methods 5b).

Liquid chromatography – tandem mass spectrometry (LC-MS/MS)

Samples were analysed using an LTQ Orbitrap Velos instrument (Thermo Scientific) coupled to a Proxeon EASY-nLC II system (Thermo Scientific). Further details on chromatographic separation and MS data acquisition are outlined in Supplementary Methods 4.

Pre-processing of LC-MS/MS data

Raw data files produced in Excalibur software (Thermo Scientific) were processed using Proteome Discoverer (PD) V1.3 (Thermo Scientific) to determine peptide identification; the subsequent Mascot (v2.3; <http://www.matrixscience.com>) output file was used for further pre-processing and analysis (supplementary methods 5a). A script was written in R to complete the pre-processing taking into account the experimental setup described above (<http://core.brc.iop.kcl.ac.uk/software/>). We named the script Pre-processing for Relative

Quantification of LC-MS/MS data (PRQ; Supplementary Methods - PRQ). PRQ performs (1) median ratio normalisation [35], (2) calculates ratios for each peptide, (3) derives protein level data from peptide scores, (4) collects protein scores across all TMT6plex's.

The University of California San Francisco (UCSF) Memory and Aging Center cohort

Replication cohort; assessments, blood collection and processing

The replication cohort consisted of samples from 79 participants enrolled in the UCSF Alzheimer's Disease Research Center (Table 1). All subjects underwent APOE genotyping, neurological and cognitive assessments [13], as well as plasma collection and storage [36] and as previously described. Clinical diagnoses of AD, FTD and MCI were made by consensus applying standard research criteria [37,38,39]. All subjects underwent PiB-PET at Lawrence Berkeley National Laboratory on a Siemens ECAT EXACT HR PET (n=69) or Biograph Truepoint 6 PET/CT (n=10) [13]. Scans were visually rated as PiB+ or PiB- by an experienced single rater blinded to clinical and plasma data [13]. Mean 50-70 min SUVR values were extracted from frontal, parietal, cingulate and lateral temporal cortex, using mean activity in the cerebellar gray matter as the reference tissue (for details of image processing see [40]).

Immunoassay – enzyme-linked immunosorbent assay (ELISA)

Single analyte sandwich ELISA was used to quantify candidate proteins and were performed as per manufacturer's instructions (Supplementary Methods 6).

Statistical analysis

All statistical analyses were performed in R (Supplementary Methods 7). For logistic and linear regression, age, gender and presence of APOE4 allele were used as covariates. For the ELISA analysis, data outliers were excluded (± 3 standard deviations) and a fourth covariate, batch, was added. PET scanner type was added as a covariate for the UCSF data. Benjamini-Hochberg Q-values were calculated as a multiple testing correction. Details of pathway, regression and classification analyses are given in Supplementary Methods 7.

Results

LC-MS/MS performed on AIBL subjects

LC-MS/MS was performed on plasma samples from 78 AIBL subjects, whose demographics are shown in Table 1. Combining data from all MS/MS runs, we identified 4,518 unique peptide sequences that corresponded to 789 unique protein groups. PRQ was able to extract 2,319 unique TMT peptides, 1,139 MW isoforms and 379 unique protein groups (Supplementary Results 1a), which was reduced to 116 confidently annotated unique protein groups after post-PRQ data clean up; this consisted of 381 protein MW isoforms (Supplementary Results 1b).

Plasma protein markers of global PiB PET

Each protein MW isoform underwent Mann-Whitney U test and logistic regression to compare PiB⁺ and PiB⁻ groups as well as Spearman Rank Correlation (SRC) and linear regression to associate protein MW isoform levels against PiB retention as a continuous measure. This was completed for both the mean and median protein roll-up methods separately, giving a total of

eight statistical tests per protein. One protein MW isoform, Complement C4a, passed all eight statistical tests. A total of 69 protein MW isoforms passed at least one statistical test (uncorrected $P < 0.05$) shown in Supplementary Results 1c. Pathway analysis (Supplementary Results 2) revealed that these protein groups were over-represented for involvement in complement and coagulation cascades ($p = 3.7 \times 10^{-22}$, $q = 3.3 \times 10^{-21}$), systemic lupus erythematosus ($p = 2.65 \times 10^{-4}$, $q = 0.15$) and prion diseases ($p = 5.9 \times 10^{-3}$, $q = 0.051$). Three albumin and 15 immunoglobulin MW isoforms were removed to leave 51 protein MW isoforms associated with PiB-PET retention (Table 2).

Subsequently, 17 proteins were selected for technical replication (Figure 1). In addition to statistical evidence we also considered the candidate's relationship with amyloid and/or AD Genome-Wide Association Studies results (Supplementary Results 3). We also chose to replicate Histidine-Rich Glycoprotein, the protein most associated with NAB, but had no prior evidence for a relationship with A β .

Technical Replication

We sought to translate our discovery findings to a simple-to-use commercially available ELISA format. The 17 proteins candidates from MS were measured in plasma samples from the 78 AIBL subjects in the discovery cohort. Using linear regression models (including age/gender/APOE/ELISA plate as covariates) we found that two proteins - α -2m ($q = 0.076$) and FGG ($q = 0.076$), replicated our findings from the LC-MS/MS discovery study (Table 3). In the discovery study, FHR-1 was increased in the PiB+ group. Although FHR-1 ($q = 0.076$) was associated with NAB at 0.1 q -value in the ELISA technical replication, an opposite trend was

observed. Apolipoprotein A-IV, Gelsolin, Histidine-Rich Glycoprotein, Haptoglobin and Apolipoprotein(a) all showed the same directional change as in the LC-MS/MS discovery.

Independent Replication

To verify the results from the AIBL samples, we measured the levels of the three proteins significantly associated with NAB (α -2m, FHR-1 and FGG) using samples from an independent cohort. These proteins were measured by ELISA in 79 samples from the UCSF cohort (Table 1). Table 4 shows that FGG was found to be significantly associated with PiB positivity, as determined both by visual examination of PiB-PET scans ($q = 5.9 \times 10^{-3}$) and by applying a threshold of 1.3 to SUVRs ($q = 0.051$). Despite not being significantly associated with NAB, α -2m correlated with SUVR positivity in the same direction as in the discovery study.

Multivariate analysis

Subjects with any missing covariates or protein measurements were excluded from the multivariate analysis, leaving 58 subjects from AIBL (28 PiB-, 30 PiB+ based on SUVR > 1.3) and 78 subjects from UCSF (46 PiB-, 32 PiB+ based on visual inspection). Classification models were trained in the AIBL ELISA data to predict SUVR positivity (> 1.3), and tested in the UCSF ELISA data to predict PiB positivity determined by visual inspection (more robust across multiple scanners). A ‘basic’ model (age/gender/APOE4) was compared to a ‘basic + proteins’ model which also used the plasma concentration of FGG, α -2m and FHR-1. Figure 2a and 2b shows a Receiver Operator Characteristic (ROC) analysis, where Area Under the Curve (AUC) was shown to be higher for the ‘basic + protein’ model than for the ‘basic’ model in the test datasets. The highest test AUC was found using the Random Forest approach, where the ‘basic +

protein' model (AUC = 0.70) outperformed the 'basic' model (AUC = 0.46) in the test dataset. The Random Forest 'basic + proteins' model gave a test set sensitivity of 50% and specificity of 85%. Additionally, a classification tree was fitted to the 'basic + proteins' model, to provide a simpler alternative with clear thresholds. The resulting classification tree used just two variables (age/plasma FGG level; Figure 2c) and achieved a comparable AUC to the Random Forest model (AUC 0.69, sensitivity 59%, specificity 78%). In the UCSF cohort, 23 out of 25 AD subjects are PiB+, it is noteworthy that the two PiB- subjects had plasma FGG levels above the threshold (Supplementary Results 4).

Discussion

With the failure of serial amyloid based therapeutics in clinical trials compromised by inclusion of substantial numbers of participants without the target pathology [18], and with the prospect of very large trials in pre-symptomatic AD such as the A4 trial and others [5], the need for blood-based markers of NAB has never been greater. Blood-based biomarkers could be used to screen large numbers of potential participants, and only those predicted to have abnormally high NAB would be retested using CSF assays or PET scans, reducing screen failure rates. This could reduce recruitment time and costs, as well as allowing eligible subjects to be identified more readily, for example from biobanks with permission for re-contact.

This study has demonstrated that a simple blood test consisting of FGG plasma levels along with age could have some potential for predicting NAB, achieving a test set sensitivity, specificity and AUC of 59%, 78% and 69% respectively, highlighting its potential use in stratifying patients for anti-amyloid trials. This independent replication was performed in a mixed dementia cohort (UCSF), suggesting that FGG and age may also have utility for distinguishing between amyloid

and non-amyloid dementias. Additionally, because the classification model was trained in a subset of the AIBL cohort containing very few AD subjects, it is more likely that FGG will be able to predict PiB positivity in non-AD subjects. However, as the UCSF cohort contained only two cognitively normal individuals, further work will be needed to determine sensitivity and specificity in people who are cognitively normal. These measures will determine the cost saving potential of this blood test for prevention trials. Preliminary data generated from a cognitively normal cohort in our laboratory supports this (Westwood et al., data not shown). Previously, Burnham et al [31] reported a blood test that achieved 79% sensitivity and 76% specificity in an independent test set. While our sensitivity is slightly lower, this is achieved by measuring a single plasma protein compared to 6 plasma proteins in the Burnham model.

While the sensitivity and specificity of these markers for predicting NAB are not high enough to use clinically, they would be useful for enrichment of clinical trials if they performed at this level in relevant populations. The strongest case can be made for prevention trials in asymptomatic subjects because of the large expected screen failure rate (~66% or higher) when looking for individuals with elevated NAB. Due to the relatively high cost of amyloid-PET scans (~\$3,000) versus blood protein ELISAs, even a blood test without clinical utility could theoretically save millions of dollars from studies of the size of A4 (n=1,000).

APOE status is a substantial risk factor for AD [41] and amyloid [30,42]. While we took APOE4 into account during our analyses we were not surprised to find that APOE genotype markers did not improve our classification model as the study was designed to be independent of this effect. However, in a general population sample APOE genotype is likely to contribute to the prediction of NAB.

It is interesting that FGG, and to a lesser extent Complement C3 and Fibrinogen α chain, were associated with NAB in this study, as this has been previously found [29,30]. However in the study by Burnham et al [31] total fibrinogen was not found to associate with NAB, whereas Kiddle et al [30] showed it was negatively associated with NAB. Further to this, decreased levels of plasma FGG have been shown to be associated with a smaller whole brain volume in AD subjects [29] whereas measures of whole fibrinogen in plasma have shown an increase [43,44]. Discrepancies of these findings may be due to the platform used to measure total fibrinogen or highlight the importance of looking at specific fibrinogen chains.

FGG is normally rejected from the brain by the blood brain barrier (BBB), yet has still been detected in mice and human brain tissue [45,46]. This could be due to the reported dysfunction of the BBB in mice [47] and humans in AD [48]. However, the movement of fibrinogen across a defected BBB seems to be molecule-specific, as smaller molecules are not BBB-permeable in AD [49]. Fibrinogen has been shown to accumulate over time as AD pathology progresses [46] and co-deposits with $A\beta$ in brain tissue [50]. Ahn and colleagues [51] demonstrated that fibrinogen binds to $A\beta$, which enhances aggregation and increases $A\beta$ fibrillisation. It is possible that decreased FGG levels associated with high NAB in our study is due to movement of fibrinogen across a compromised BBB in subjects with AD pathology.

After FGG, α -2m was the second most promising candidate, shown for the first time to associate with NAB. This is noteworthy because, α -2m has been found to be one of the most replicable markers of other AD-related phenotypes including diagnosis, hippocampal metabolism and response to treatment with divalproex sodium [20]. Future studies should aim to replicate all

previously discovered markers of NAB and investigate which combination of analytes would achieve higher sensitivity and specificity.

To our knowledge, this is the first study to apply an unbiased and non-targeted quantitative LC-MS/MS discovery approach, combining LC-MS/MS with TMT-labelling, for the investigation of plasma proteins related to NAB. Furthermore, this method will allow the unprecedented exploration of plasma peptide and modified proteins as markers of NAB. We also describe a novel and automated bioinformatic pipeline - PRQ - to accurately pre-process TMT-MS data. PRQ not only conducts rigorous normalisation of MS data [35] but also automates the calculation of peptide/protein ratios against the study reference.

Subsequently, technical replication was performed to reduce the number of false positives and to ensure translation of LC/MS-MS findings using a platform more applicable to clinical setting. Using commercially available immunoassays, we confirmed that α -2m, FGG and FHR-1 significantly predicted NAB with a 0.1 Q-value significance level. All except FHR-1 displayed a similar direction of association between discovery and replication. Immunoassays cannot always distinguish between sequence variants, proteins modified with different PTM, or different truncated forms of a same protein seen by LC/MS-MS. This could also explain the differences seen in association trend between discovery and replication in some cases, e.g. FHR-1; therefore these candidates should not necessarily be discounted. The discrepancies observed between the two platforms point to the need of investigating protein modifications as potential biomarkers in future studies.

The discrepancies between findings in AIBL and UCSF could be due to low statistical power, differences in disease stage or pre-analytical factors. The major difference in pre-analytical

factors is the centrifugation step of plasma collection: AIBL has a two-step centrifugation (200 x g, remove supernatant, then 800 x g), whereas UCSF has a single centrifugation step (1300 – 1800 x g). This highlights the importance of standardization of blood collection and preparation for biomarker studies.

While many agree that A β deposition is the earliest event in AD pathogenesis, one group has shown changes in episodic memory preceding changes in A β levels [52]. If confirmed in other cohorts it would be interesting to compare the ability of episodic memory and our blood test to predict NAB in asymptomatic individuals.

In summary, the current study presents a potential blood test, consisting of measuring FGG, which along with age has some ability to predict NAB in an independent sample. To ensure robustness and relevance of these findings, this test will need to be replicated in larger cohorts that are more representative of relevant clinical trial populations. This study adds further evidence that differences in the plasma proteome in relation to AD and its pathology do exist, and therefore such changes could be used to stratify patients for anti-amyloid treatment trials. This could lower barriers to the development of an effective treatment to combat the increasing concern of dementia.

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Table 1 – Demographics of selected subjects from the AIBL and UCSF cohorts.

	AIBL Discovery Cohort			UCSF Replication Cohort		
	Low Neocortical SUVR (PiB-)	High Neocortical SUVR (PiB+)	P-value	Low Neocortical Visual PiB read (PiB -)	High Neocortical Visual PiB read (PiB +)	P-value
Number of subjects (<i>n</i>)	38	40		47	32	
SUVR (mean (s.d.) {missing})	1.11 (0.06)	2.34 (0.33)	2.4×10^{-25}	1.2 (0.12) {1}	2.2 (0.35) {2}	4.2×10^{-16}
Gender; females (<i>n</i> (%))	18 (47%)	20 (50%)	0.83	18 (38%)	14 (44%)	0.65
Age in years (mean (s.d.))	75.8 (6.53)	80.9 (8.22)	0.0035	65 (8.8)	64 (8.4)	0.61
Clinical diagnosis (<i>n</i> (%))	HC: 13 (34%)	HC: 6 (15%)	0.0037	HC: 2 (4.3%)	HC: 1 (3.1%)	1.9×10^{-10}
	SMC: 18 (47%)	SMC: 13 (40%)		MCI: 1 (2.1%)	MCI: 1 (3.1%)	
	MCI: 7 (19%)	MCI: 16 (30%)		AD: 2 (4.3%)	AD: 23 (72%)	
	AD: 0 (0%)	AD: 6 (15%)		FTD: 42 (89.3%)	FTD: 7 (21.8%)	
APOE ϵ 4 carrier (<i>n</i> (%))	14 (37)	25 (63)	0.36	8 (17%)	13 (41%)	0.036
MMSE (mean (s.d.))	28.3 (1.8)	26.8 (4.1)	0.038	26 (4.3)	21 (6.9)	0.0011

Table 2 – LC-MS/MS data; Protein MW isoforms significantly associated with NAB. (All multiple testing corrected Q-values were > 0.75). For regressions age, gender and presence of APOE4 was used as covariates.

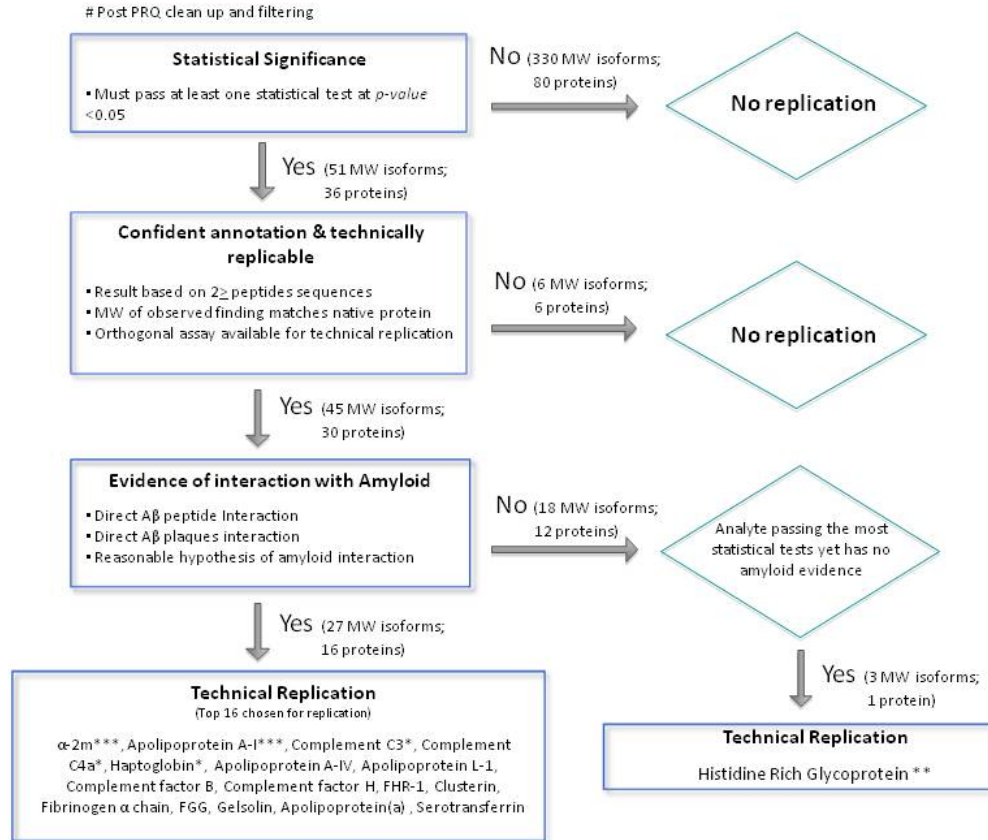


Figure 1 – Flow diagram to select LC-MS/MS plasma NAB candidate markers for technical replication. * Two protein MW isoforms associated with NAB; ** Three protein MW isoforms associated with NAB; *** Four protein MW isoforms.

Table 3 – Technical replication of plasma protein candidates discovered by LC-MS/MS.

UniProt ID	Protein Name	Number of outliers excluded	Logistic regression with SUVR > 1.3			Linear regression with SUVR		
			Beta	P-value	Q-value	Beta	P-value	Q-value
P01023	α-2-macroglobulin (α-2m)	10	1	8.9x10 ⁻³	0.076	0.2	7.9x10 ⁻³	0.068
Q03591	FHR-1	11	-1	4.6x10 ⁻³	0.076	-0.22	5.5x10 ⁻³	0.068
P02679	Fibrinogen γ chain (FGG)	0	-0.7	0.041	0.23	-0.2	0.014	0.081
P08519	Apolipoprotein(a)	21	0.48	0.13	0.34	0.18	0.042	0.18
P06396	Gelsolin	2	-0.48	0.11	0.34	-0.14	0.068	0.19
P00738	Haptoglobin	2	-0.38	0.18	0.39	-0.13	0.089	0.19

P04196	Histidine Rich Glycoprotein	2	0.48	0.14	0.34	0.14	0.081	0.19
P06727	Apolipoprotein A-IV	2	-0.63	0.083	0.34	-0.17	0.067	0.19
P01024	Complement C3	0	-0.61	0.25	0.47	-0.21	0.13	0.25
P0C0L4	Complement C4a	0	-0.55	0.51	0.66	-0.27	0.22	0.38
P10909	Clusterin	0	-0.27	0.36	0.51	-0.091	0.27	0.41
P02647	Apolipoprotein A-I	0	0.34	0.29	0.47	0.088	0.32	0.46
P02671	Fibrinogen α chain	6	-0.28	0.3	0.47	-0.064	0.39	0.52
P02787	Serotransferrin	1	-0.013	0.96	0.96	-0.041	0.6	0.73
O14791	Apolipoprotein L-1	0	-0.09	0.74	0.89	-0.026	0.73	0.77
P08603	Complement factor H	3	0.066	0.8	0.89	0.027	0.7	0.77
P00751	Complement factor B	2	0.053	0.84	0.89	0.018	0.81	0.81

Table 4 – Independent replication of plasma protein candidates discovered by LC-MS/MS and technically replicated. Only one outlier (> 3 standard deviations from mean) was excluded, which was detected for FGG. For regressions age, gender, presence of APOE4, ELISA plate and scanner type were used as covariates.

UniProt ID	Protein Name	Logistic regression to visual read			Logistic regression to SUVR > 1.3			Linear regression to SUVR		
		Beta	<i>P</i> -value	<i>Q</i> -value	Beta	<i>P</i> -value	<i>Q</i> -value	Beta	<i>P</i> -value	<i>Q</i> -value
P01023	α -2m	-0.013	0.96	0.96	0.27	0.29	0.44	0.075	0.22	0.33
P02679	FGG	-1.0	2.0×10^{-3}	5.9×10^{-3}	-0.74	0.017	0.051	-0.21	4.1×10^{-4}	1.2×10^{-3}
Q03591	FHR-1	-0.066	0.79	0.96	0.011	0.97	0.97	1.5×10^{-3}	0.98	0.98

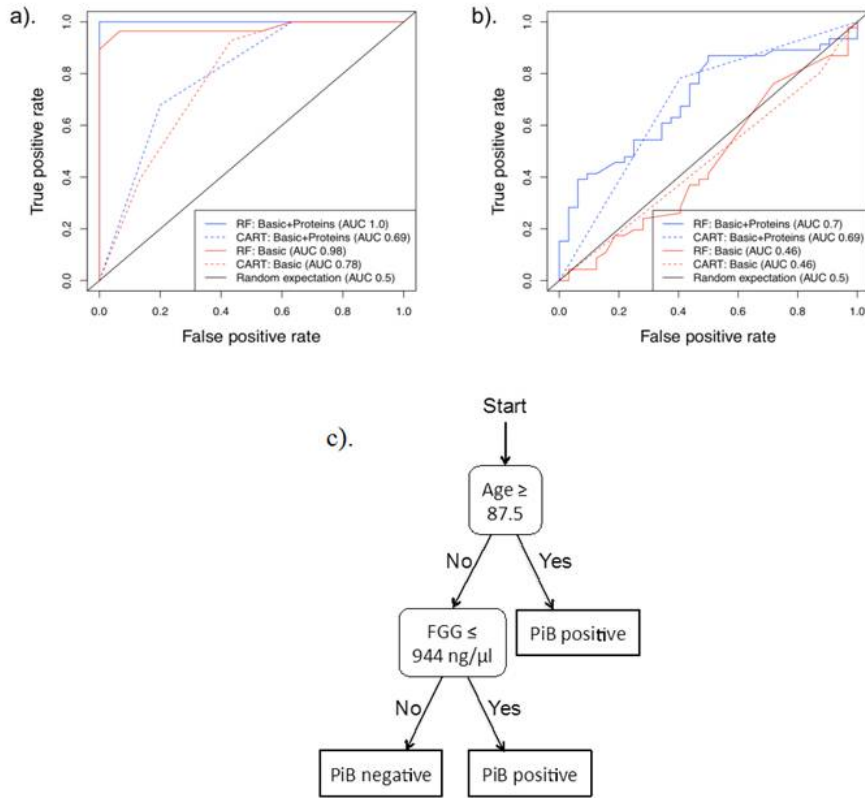


Figure 2 – Receiver Operator Characteristic (ROC) curves for the prediction of PiB positivity. A ‘basic’ model (age/gender/APOE4 presence) is compared to a ‘basic + proteins’ model also including the plasma levels of FGG, α -2M and FHR-1. Random Forest (RF) and Classification and Regression Trees (CART) were used to fit models in CARET using default parameters. Area Under the Curve (AUC) is given for each model. ROC curves are shown comparing predictive accuracy of models in (a) the training dataset (AIBL), and (b) the test dataset (UCSF). Classification tree trained on AIBL ELISA data to predict NAB positivity and estimated cut-off (c).