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Blood-based biomarkers for Alzheimer's disease: mapping the road to the clinic

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Abstract

Biomarker discovery and development for clinical research, diagnostics and therapy monitoring in clinical trials have advanced rapidly in key areas of medicine, notably oncology and cardiovascular diseases, allowing for rapid early detection and supporting the evolution of biomarker-guided precision medicine-based targeted therapies.

In Alzheimer's disease (AD), breakthroughs in biomarker identification and validation include the cerebrospinal fluid (CSF) and positron emission tomography (PET) markers of amyloid β (A β) and tau proteins, which are highly accurate in detecting the presence of pathophysiological and

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neuropathological changes of AD. However, their high cost, insufficient accessibility, or invasiveness may limit their use as viable first-line tools for detecting patterns of pathophysiology. Therefore, a multi-stage, tiered approach is needed, prioritizing development of an initial screen to exclude from these tests the high numbers of people with cognitive deficits who do not demonstrate evidence of underlying AD pathophysiology.

This perspective summarizes the efforts of a working group that aimed to survey the current landscape of blood-based AD biomarkers, and outlines operational steps for an effective academic-industry co-development and path forward from identification and assay development to validation for clinical use.

Introduction

Alzheimer's disease

Alzheimer's disease (AD) is a clinically and pathophysiologically heterogeneous complex neurodegenerative disease (ND). AD is the most common cause of age-related ND, impacting millions of individuals worldwide; currently, one out of nine people over the age of 65 are living with AD¹ and the prevalence of AD is expected to grow exponentially over the next several decades¹.

The pathogenesis of AD involves interacting pathophysiological cascades, including core events, i.e. accumulation of the 42-amino acid-long amyloid beta peptide ($A\beta_{1-42}$) into amyloid plaques in the brain parenchyma and the formation of intraneuronal neurofibrillary tangles composed of hyperphosphorylated tau protein². Emerging evidence stresses the existence of additional molecular pathophysiological pathways, such as axonal disintegration³, synaptic dysfunction and degeneration⁴, innate immune response and neuroinflammation^{5,6}, vascular and cell membrane dysregulation⁷, and brain metabolic dysfunction⁸ – across the different stages of AD. Moreover, other proteinopathies and pathologies frequently co-exist in the ageing brain. These include TDP-43 or α -synuclein proteinopathies, non-AD tauopathies, vascular pathology, and hippocampal sclerosis^{9–12}. For these reasons, establishing a definitive diagnosis and developing effective treatments of AD is complicated.

At present, the aetiology and pathogenesis of AD is the subject of ongoing research and debate. The amyloid cascade hypothesis proposes that the brain accumulation of aggregated forms of $A\beta$ is the trigger and/or driver of the disease process¹³. However, recent studies raised questions about this hypothesis as the exclusive cause and/or intervening link between the pathophysiology of AD and its clinical phenotype. The notion that biochemical and cellular mechanisms generate complex cognitive alterations has renewed AD research, leading to replace the first descriptive studies with a molecular, mechanistic view. The exponential increase in knowledge on interacting pathogenic mechanisms in individuals suffering from AD holds promise for the development of future biomarker-guided targeted therapies and prevention strategies^{14–17}.

The potential impact of biomarkers on primary care and neurology for the detection and diagnosis of AD

Given the complex clinical phenomenology, clinical, neurological, and neuropsychological examinations are still an integral component of accurate late-stage detection of clinically symptomatic AD and other ND. However, waiting times for appointments with specialists in the U.S., U.K. and Ireland (and other countries) can be very long resulting in substantial and often critical delays for patients and providers¹⁸. Memory clinics or general neurology clinics in many countries receive a broad range of referrals covering many conditions and diseases, therefore the improved streamlining of referrals to specialty clinics can have a significant impact on health care utilization and costs¹⁸. As one example, recent U.S.-based legislation requires that elderly people aged 65 and older receive annual cognitive examinations as part of the Annual Wellness Visit (CMS.gov)¹⁹; however, older adults continue to be inadequately assessed for cognitive decline during primary care visits²⁰. Given that the average duration of primary care visits for geriatric patients is 21 minutes²¹, this is perhaps not surprising. Additionally, cognitive examinations are regularly administered, scored, and interpreted incorrectly in primary care due to lack of training and expertise^{22,23} and there are significant differences between primary care and specialty care approaches to diagnosis, treatment, and social support²⁴. Therefore, a process that aids primary care practitioners in deciding which patients should receive a referral to a memory clinic would be of substantial advantage to both specialists and general practitioners. Such a system would reduce the overall clinic and medical system burden by decreasing the numbers of unnecessary referrals and diagnostic procedures^{25,26}. Biomarker-based diagnostics can greatly aid a multi-stage selection of patients into appropriate centres.

To date, the literature has focused on diagnostic biomarkers^{27,28} for specialty clinic settings with little attention to screening instruments required for broad-based implementation in primary care settings. Additionally, diagnostic paradigms continue to rely substantially on clinical symptoms, with only relatively recent research guidelines taking biomarkers of AD pathophysiology into account^{29,30}. Up to the mid-2000s, diagnosis of AD had a “clinicopathological” basis, with no definitive positive diagnosis possible until post-mortem confirmation of the presence of A β plaques and neurofibrillary tangles, and clinical diagnosis being assigned after a patient had reached the late-stage syndromal dementia threshold, by exclusion of other aetiologies of dementia²⁹. This concept of late-stage diagnosis of exclusion focusing on clinical phenotype showed profound limitations, given that clinical symptoms are by nature heterogeneous (including an increasing number of atypical clinical phenotypes) and vary regarding onset, progression, and sequence of events. They have considerable overlap with other central nervous system (CNS) proteinopathies causing dementia as well as with dementia due to cerebrovascular disease³¹. More recently, with the advent of reliable radiotracers with affinity for cerebral amyloid deposits³², and the availability of clinically well-validated assays for cerebrospinal fluid (CSF) detection of A β peptides and tau proteins³, there has been progress towards a “clinicobiological” diagnostic approach in which pathophysiological and topographic biomarkers are integrated into the diagnostic paradigm²⁹ as adjuncts to core clinical criteria³⁰. These biomarkers are clinically available in specialty clinic settings for some countries, providing the means for dementia specialty clinicians to confirm diagnosis with much higher certainty through the presence of

characteristic features of AD pathophysiology, thereby allowing for a more conclusive aetiological diagnosis^{14,33,34}. Nevertheless, use of these technologies is not yet the standard and they are restricted to specialty clinic settings. On the other hand, biomarkers are now being incorporated into the comprehensive clinical diagnostic process, which will result in improved overall accuracy in detection of pathophysiology. However, as was the case for more matured translational research areas of biomedicine, such as oncology and cardiovascular medicine, the field is in need of biomarker-based diagnostics that can accurately and reliably identify individuals at risk and patients as precise and early as possible in the disease process^{35–39}, preferentially in primary care settings. The availability of these biomarker-based diagnostics (e.g. CSF A β _{1–42} and tau) is also expected to open the window for precision medicine and allow a shift away from the traditional “one-size-fits-all” approach with “magic bullet” drugs in neuroscience to the development of biomarker-guided targeted therapies^{25,40–43}.

Facilitation of the precision medicine paradigm in ND such as AD requires an array of converging advanced analysis methods. Through systems theory, it is possible to conceptualize original and innovative models to explicate all systems levels – assessed *via* systems biology and systems neurophysiology⁴² – and different data dimensions in space and time of the non-linear, dynamic, and chronically progressive nature of the clinically and biologically heterogeneous construct of polygenic AD^{35,44}, historically developed after the clinical description of the first patients and the associated brain histopathological remarks⁴⁵.

The agnostic, hypothesis-free systems theory approach seems appropriate to explain the complex and heterogeneous origin and time course of the pathophysiological failure underlying the different forms of AD³⁵. For multifactorial diseases, comprehensive holistic systems-level approaches are required; this is the case of the systems biology model aiming at understanding the genotype-phenotype relationships and the mechanisms at the level of the genome/epigenome, transcriptome, microRNome, proteome/peptidome, metabolome/lipidome, microbiome, and environmental factors participating in complex cellular networks^{35,44,46}. Longitudinal investigations using systems biology-based strategies help characterize the complex molecular pathophysiology of the subforms of AD, all evolving through the convergence of alterations of homeostasis and/or failures in many systems, networks, signalling pathways, or pathophysiological processes⁴⁴.

One of the key objectives of precision medicine is to bring new models for prevention, early detection, differential diagnosis, and treatment of AD and other ND, according to individual biological differences reflected by multimodal biomarkers^{33,35,37}. Therefore, following the advanced models of oncology and cardiovascular medicine, innovative biomarker studies are expected to detect specific diagnostic, prognostic, and predictive biomarker signatures to adapt the therapy to individual patients⁴⁰. As reported by the Institute of Medicine (IOM) Committee Recommendations for Advancing Appropriate Use of Biomarker Tests (companion diagnostics) for Molecularly Targeted Therapies, the ultimate goal of precision medicine is to enhance both clinical outcomes and the quality of patient care⁴⁷.

Precision medicine is being used in AD and in a growing number of ND as a result of: 1) the development of large-scale biological databases and 2) the evolution of advanced high-

throughput “omics” sciences. Novel methods developed within the “omics” disciplines have been successfully applied in the more advanced cancer area and are expected to transfertilize to AD and other ND. For instance, validated microarray expression profiling and current RNAseq technologies allow identifying differential expression of whole genomes in any specific sample at any given time point. Numerous reports exploiting these transcriptomics methods to recognize biomarkers in specific cancer subtypes are present⁴⁸. The participation of microRNAs (miRNAs) in crucial cellular processes including cell death and their role in negative control of the expression of various oncoproteins make them interesting candidate biomarkers for cancer. Cancer-specific markers are detected in the blood from the earlier stages of tumor development. Since their concentration increases as tumor advances over time, they are useful dynamic indicators of tumor growth⁴⁹. In proteomics, innovative methods of immuno-PCR, based on conjugations of specific antibodies and nucleic acids and exploiting ultrasensitive PCR signal amplification, can result into a 100 to 10,000-fold increase in sensitivity compared with the analogous enzyme-amplified immunoassays, hence considerably increasing the sensitivity of protein biomarker detection^{50,51}. Powerful computational and integrative network biology tools are needed to assimilate the large multimodal information generated by the “omics” exploratory analyses and to comprehensively understand the decompensations at systems level⁵². An advantage in the AD field is that many of the pathologies are well known. Hence hypothesis-free systems theory approaches may be advanced in synergy with the refinement of methods that directly reflect core pathologies, much like what has recently happened in the field of blood-based direct, A β -related biomarkers for amyloid pathology^{53–55} and blood biomarkers for axonal degeneration^{56,57}.

Why are blood-based biomarkers needed for AD?

While advancements in positron emission tomography (PET) and CSF biomarker analyses have the potential to improve accuracy within the diagnostic process, these methods show limitations precluding their use as first-line diagnostic tools. These limitations would be countered by use of blood-based biomarkers²⁸. If a sufficiently accurate and standardized blood test can be developed, it is highly likely to be more cost effective than PET scans, the substantially high cost of which limits accessibility, generalizability and availability^{58,59}, particularly outside the U.S.A. Because of this high cost, PET testing in AD will likely only become reimbursed as an adjunct to other less expensive tests, as was previously the case with PET scans for cancer. In addition, blood testing is less invasive than CSF testing, which requires lumbar puncture^{60,61}. Furthermore, blood testing is already a well-established part of clinical routines globally, requiring no further introduction and training for health care professionals (HCPs), and can be easily performed in a variety of relevant settings (including primary care, in community-based medicine centres or even in a patient’s home) and at repeated intervals⁶². Blood sample handling infrastructure is also well established, allowing for the shipping of samples during the initial development stages. Therefore, blood-based screening biomarkers for AD can meet the scalability needs required for primary care settings and even for the broad population-based screening approach that may evolve with the advancing innovative precision medicine framework. Finally, the use of blood-based biomarkers offers the potential for testing of a huge range of comprehensive exploratory and candidate pathophysiological biomarkers, reflecting the full *spectrum* of disease triggering

and driving molecular mechanisms underlying polygenic AD, beyond the conventional amyloid- and tau-based tests.

Blood-based biomarkers are an ideal choice as the first-step of the multi-stage diagnostic process beginning in primary care settings, and provide the means to determine which individuals or patients should receive referral to assessment by specialists, including diagnostic CSF analysis, magnetic resonance imaging (MRI) or amyloid PET diagnostics^{26,63}, which is analogous to the existing process for many other conditions. In addition to meeting the significant clinical need, the availability of these tools would provide a viable path towards regulatory⁶⁴ and reimbursement approval, using the cancer paradigm as a model.

Physiological challenges of developing blood-based biomarkers for AD

Identifying potential blood-based biomarkers for CNS diseases offers several challenges (Figure 1). First of all, the vast complexity of blood, which contains both cells and extracellular fluid, needs to be considered. Blood includes a range of different molecules (proteins, peptides, nucleic acids, lipids, metabolites) that can be detected in plasma, exosomes, and cellular compartments. The latter includes erythrocytes, leucocytes, and platelets, isolated into distinct cell subsets *via* flow cytometry or *via* buffy coat after density gradient centrifugation. Given the presence of different unique cellular compartments, each one is a potential source of biomarkers and may introduce variability to analyses^{62,65}. The diversity of blood candidate biomarkers is relevant and includes: 1) protein concentrations/activity/isoforms and post-translational modifications; 2) metabolic products, such as amino acids, carbohydrates, lipids, organic acids; 3) nucleic acids. In the latter, single nucleotide polymorphisms (SNPs) – DNA sequence variations occurring when a single nucleotide in the genome differs between members of a species – can act as biological markers, helping localize genes associated with the disease⁶⁶. When SNPs occur within a gene or in a regulatory region adjacent to a gene, they may affect the gene's function, thus playing a key role in disease development. The growing application of high-throughput next-generation DNA sequencing technologies is playing a substantial role in screening whole genomes and, consequently, in identifying novel genetic variants affecting the risk of AD^{67,68}. The 1000 Genomes Project^{69,70}, supported by the U.S. National Human Genome Research Institute consortium, made significant progresses toward this aim.

Secondly, the CNS is an effectively contained environment, and potential biomarkers may be present at very low concentrations in blood once they have crossed the blood–brain barrier (if they cross it at all as intact molecules)^{59,71}. Indeed, there is minimal evidence of the peripheral effects of AD⁷², although there is some evidence that the blood–brain barrier may be increasingly compromised in normal aging and with increased AD progression^{73,74}. Also, physiological mechanisms occurring peripherally may hamper the clinical utility of blood-based AD biomarkers, e.g. acute-phase or inflammatory proteins and small molecules, and metabolites present also in peripheral organs⁵⁹. In particular, possible confounders include: significant biomarker dilution, considering the modest volume of the CSF and the extensive volume of the blood and extracellular fluid⁷⁵; degradation in the liver or directly in the blood by proteases; matrix effects caused by adherence to plasma proteins or even

blood cells; and excretion from the kidneys. These factors may substantially lower the concentrations of the biomarker, as well as decrease the time window for testing. A further major issue is the frequent existence of overlapping ND and co-morbidities in patients with AD, including cardiovascular, respiratory, hepatic, renal, and rheumatic disease, all of which may affect protein profiles in plasma⁵⁹. On the other hand, since AD is a complex polygenic disease, amyloid and/or tau accumulation never occur in isolation of other relevant molecular/cellular pathophysiological mechanisms and brain systems failures in disease models. Elucidating this true complexity and heterogeneity, including through blood-based biomarker analyses, may substantially aid in a more comprehensive understanding of the disease for the generation of the precision medicine paradigm^{25,40,44}.

Gathering consensus on blood-based biomarker development in AD

Given the strong rationale yet current challenges for the development of a blood-based biomarker for AD, an international, interdisciplinary expert working group was convened in March 2016 to discuss the ideal development process. The meeting was supported by Roche Diagnostics International. The working group was selected based on profiling of global experts and working groups that was done within the extensive biomarker literature search. Experts were selected based on their publications and involvement in research in the field of neurodegenerative biomarkers in CSF and in blood. The nature of support of Roche Diagnostics International was the assistance of organizing of scientific meetings, video, and teleconferences based on the scientific chair guidance and the support of the scientific writing agency. The meeting room as well as refreshments according to the guidance of the compliance guidelines were provided.

Process of the working group—Prior to the working group meeting, an in-depth comprehensive review of the literature was conducted (including “grey literature” – patents, press releases, and proprietary databases) to identify candidate biomarkers (see “Current landscape” below for further details). A structured and detailed pre-work survey of the meeting expert attendees provided additional material for analysis of the current blood-based biomarker landscape in AD and the most pressing challenges for the successful development of further markers. The aim of the survey was to identify and rank candidate blood-based biomarkers, which were selected from the landscape analysis, in order of priority for further development.

The working group itself provided a forum for definition of an ideal target product profile for a blood-based biomarker in AD. A critical appraisal of the selected biomarkers was then performed to determine if they met this ideal profile, and to assess the general quality of research findings and their suitability for further development. Further considerations were the optimal validation process for a biomarker, the best route to determine the efficacy of a biomarker (predictive values versus accuracy) and the ideal cohorts in which biomarkers could be evaluated. In-depth discussions were also held on several candidate biomarkers, with the aim of defining key questions critical for evaluating any potential current or future candidate.

Current landscape

A landscape analysis was conducted to review all blood-based biomarkers in development for AD globally, and then assess these biomarkers on the basis of pre-defined selection criteria in order to prioritize those with the greatest likelihood of successful implementation. The data were sourced from publications (MEDLINE® database; 2011–2015), patents (worldwide patents, including PCT, USA and China; 2008–2015); press releases and conference abstracts from eight major neuroscience or AD conferences; and ChinaBio® proprietary databases.

After screening more than 28,000 reports, 1,404 studies of blood-based biomarkers for AD were identified, 1,039 of which were from publications and conference abstracts. The 1,039 studies were categorized according to biomarker type; 18% of the studies were on the conventional AD biomarkers, A β and tau (or their associated peptides/proteins or variants), 19% were on genetic markers, 29% were on biomarker panels, and 34% were on markers related to emerging mechanisms such as inflammation, immune response, oxidative stress, DNA damage, mitochondrial dysfunction, and neuronal or microvascular injury. The identified biomarkers were then screened regarding intended use, asset type, technology platform, test type and analyte, and whether they included human/clinical data. A final screening was performed to focus only on high-quality candidates, according to novelty, cohort quality, presence and quality of validation study, diagnostic performance, and perceived quality of research. The final list contained 196 candidate biomarkers, with biomarker panels and emerging targets being the most common categories (Figure 2A). Immunoassays and molecular assays were the most popular technology platforms utilized for high-quality AD blood-based biomarkers, accounting for more than 70% of studies (Figure 2B). A potential limitation of the search was the relatively narrow date range, which may have resulted in the exclusion of some potentially useful biomarkers that were published outside of this date range. The most promising biomarker candidates and the key studies, including those published since the original analysis, are summarized below.

Among the conventional AD biomarkers, the ratio of A β _{1–42}/A β _{1–40} has shown potential as a screening/diagnostic marker in several studies. Most early studies on plasma A β _{1–42}, A β _{1–40}, and A β _{1–42}/A β _{1–40} using enzyme-linked immunosorbent assay (ELISA) methods, found no or minor differences between AD and control groups^{76–78}. More recently, a large cohort study using a novel ultrasensitive immunoassay technique (single molecule array (Simoa)), showed weak but significant correlations between plasma and CSF levels of A β _{1–42}, A β _{1–40}, and A β _{1–42}/A β _{1–40}⁵⁵. In addition, plasma levels of A β _{1–42}, A β _{1–40}, and A β _{1–42}/A β _{1–40} were lower in AD patients compared with cognitively healthy, mild cognitive impairment (MCI) and subjective cognitive decline (SCD) subjects⁵⁵. An alternative technique for measuring A β peptides in plasma involves immunoprecipitation and mass spectrometry^{54,79}. A pilot study found a trend for a reduction in both plasma A β _{1–42} and A β _{1–42}/A β _{1–40} in AD compared with controls⁷⁹. A separate study, using a modified method involving proteolytic digestion of A β peptides before mass spectrometry, found that plasma levels of A β _{1–42} and A β _{1–42}/A β _{1–40} were reduced in amyloid PET-positive subjects⁵⁴. Furthermore, plasma A β _{1–42}/A β _{1–40} had a good diagnostic accuracy, as indicated by a receiver operating characteristic (ROC) area under the curve (AUC) value of 0.8865. A

recent study further supports the use of immunoprecipitation–mass spectrometry methods to measure plasma A β peptides⁵³. Nakamura and colleagues reported high performance of the ratios of plasma A β _{1–40}/A β _{1–42} and amyloid precursor protein (APP)_{669–711}/A β _{1–42}, and their composite, for the prediction of amyloid PET burden⁵³.

Results in the field have been mixed as Lewczuk and colleagues did not demonstrate plasma A β _{1–40}/A β _{1–42} to be a useful diagnostic in a multi-center study⁸⁰. Prior work also suggests that plasma amyloid markers may be impacted by cardiovascular and cerebrovascular factors^{81,82} and may impact the diagnostic and predictive of these markers⁸³. Therefore, while promising, additional work is needed to determine if plasma A β has utility as a screening tool for brain amyloidosis and AD, including large clinical and prospective cohorts as well as direct comparisons of different pre-analytic and analytic methodologies since immunoprecipitation–mass spectrometry could be difficult to generalize for clinical use⁸⁴.

Plasma levels of tau protein have also been successfully quantified using novel, highly sensitive immunoassay techniques, and different technologies have shown plasma tau to be increased in AD compared with controls^{85,86}. Plasma tau levels showed a strong association with AD in a meta-analysis (average ratio of levels in AD versus controls: 1.95, 95% confidence interval 1.12–3.38, $p=0.02$)³. A study based on the large Alzheimer's Disease Neuroimaging Initiative (ADNI) and Biomarkers For Identifying Neurodegenerative Disorders Early and Reliably (BioFINDER) cohorts confirmed an increase in plasma tau in AD dementia, but with a substantial overlap in levels with cognitively unimpaired elderly subjects⁸⁷. Interestingly, longitudinal evaluations showed correlations between high baseline levels of plasma tau levels and future cognitive decline, increased atrophy rates (measured by MRI) and hypometabolism (measured by ¹⁸F-fluorodeoxyglucose-PET (¹⁸F-FDG-PET))⁸⁷. Thus, current data suggest a minor increase in plasma tau in AD, although with too large overlap with controls to be diagnostically useful. In a recent study, elevated plasma tau levels were associated with cognitive decline, independent of elevated brain A β ⁸⁸, suggesting potential use as a non-disease specific screening tool.

Among the emerging targets category is the axonal protein, neurofilament light (NF-L), which can be quantified using the ultrasensitive Simoa technique⁸⁹. Serum NF-L levels are highly correlated with CSF levels, suggesting that blood measurements reflect brain pathophysiology⁹⁰. A recent study on the ADNI cohort found a marked increase in plasma NF-L in AD cases with a ROC AUC value of 0.8757, which is comparable to the plasma A β _{1–42}/A β _{1–40} ratio results reviewed above. Plasma NF-L was highest in MCI cases with positive amyloid PET scans, and predicted faster cognitive deterioration, higher rate of both future brain atrophy (measured by MRI) and hypometabolism (measured by ¹⁸F-FDG-PET)⁵⁷. Interestingly, a study on familial AD (FAD) showed that blood NF-L was increased not only in symptomatic FAD cases, but also in pre-symptomatic mutation carriers with levels correlating with expected estimated year of symptom onset as well as cognitive and MRI measures of disease stage⁵⁶. These findings suggest that plasma NF-L detects neurodegeneration in the preclinical phase of AD. However, high plasma (or CSF) NF-L is not specific for AD, but found in several neurodegenerative disorders such as frontotemporal dementia, progressive supranuclear palsy and corticobasal syndrome^{91–93}. Thus, plasma

NF-L might have an application as a screening test for neurodegeneration in the initial primary care evaluation of patients with cognitive disturbances.

Also in the emerging targets category is β -site amyloid precursor protein cleaving enzyme 1 (BACE1), the β -secretase responsible for the first cleavage step required for the generation of A β peptides from APP⁹⁴. Studies using ELISA-based methods have shown increased BACE1 activity in the plasma of AD patients compared with controls^{95,96}. A recent study found that plasma BACE1 activity was increased in both MCI subjects and AD patients compared with healthy controls, and significantly higher in MCI subjects who progressed to AD (over 3 years' follow-up) than in cognitively stable MCI subjects who did not progress⁹⁶. These results suggest that plasma BACE1 activity has potential as a biomarker to predict progression from MCI to AD dementia, which could be valuable in the primary care and clinical trials settings; however, further studies are needed to validate these findings.

These four candidate blood-based biomarkers (A β_{1-42} /A β_{1-40} , tau, NF-L, and BACE1) and their roles in AD pathogenesis are depicted in Figure 3.

Blood-based biomarker panels are another area of interest for AD as a combination of markers might show better separation between groups than single biomarkers. In recent years, several protein panels have shown diagnostic or prognostic potential^{63,97-99}, including a 21-protein panel for AD screening, which has demonstrated a positive predictive value (PPV) of 0.85 and a negative predictive value (NPV) of 0.94 in a preliminary validation⁶³. Non-protein analytes such as amino acid¹⁰⁰, miRNA^{101,102}, and lipid panels¹⁰³ have also shown promise but independent, large-scale validation studies are needed.

The road to the clinic

Historical difficulties with development

Assay reliability, and robust replication and validation of initial results are key issues for blood-based biomarker development^{26,71,104,105}. Historically, an important limitation has also been the analytical sensitivity of available technologies; standard immunochemical methods have often not been sensitive enough to allow for the reliable quantification of CNS-derived molecules in the blood, a situation that has changed with the advent of ultrasensitive assays¹⁰⁶. There is considerable variability between studies owing to inconsistencies in clinical cohorts (e.g. diagnostic evaluations, stage of disease), sample availability, difficulties in standardizing the samples themselves, and pre-analytical and analytical differences. Similar to CSF assays, consistent use of automated assays between studies may ameliorate analytical differences. However, the lack of standardization of pre-analytical protocols is also a major issue. For example, it is known that the majority of errors in proteomic analysis arise in the pre-analytical phase^{65,107}. Pre-analytical variables can be divided into several main categories (Table 1)^{71,107}. Initial steps towards standardization have already been made¹⁰⁷, with the formation of a professional interest area focused on Biofluid Based Biomarkers (BBB-PIA) (in association with the Alzheimer's Association's International Society to Advance Alzheimer's Research and Treatment [ISTAART]). This

group will drive towards consensus on harmonization of pre-analytical and analytical protocols, and will lead the development of a repository of clinical reference samples to enable assay harmonization and clinical performance assessment. However, further action is needed and communication of these issues to wider audiences is also warranted.

The experimental design of a biomarker evaluation study (including obtaining relevant cohorts with appropriate sample availability) will be important, as this has proved difficult in previous studies. Whilst many cohorts exist and potentially have samples suitable for use to develop a blood-based biomarker, these cohorts may not match the patient profile encountered in the target setting. Finally, any candidate biomarker will need to offer excellent scalability to allow for large-scale use.

Ideal characteristics of a blood-based biomarker

The first steps in development of a blood-based biomarker for AD diagnosis should be to: a) finalize the specific context of use (COU) for regulatory purposes; and b) determine the target product profile, which allows for targeted development of the biomarker to match specific requirements for the intended COU. This prospective approach may enhance success compared with current approaches based on a discovery model, where biomarkers are first identified in case-control studies before being retrospectively adapted to an intended COU. In cases where a COU is prospectively defined, the original case-control study may not be fit-for-purpose to adequately test the COU. During the working group meeting, the consensus was that the COU for a blood-based biomarker in AD should be as a tool to assess patients reporting cognitive deficits in the primary care setting, allowing for identification of the subset of patients demonstrating biological signs consistent with AD who require further specialist diagnostic testing. Ideally, the biomarker could also be used to rule out other ND with symptoms similar and/or overlapping to those of AD. Validation against current clinical and biomarker-based diagnostic paradigms would also be required (ideally the IWG-2 diagnostic criteria²⁹: evidence of cognitive impairment in conjunction with amyloid PET or the CSF “AD signature” [reduced A β and elevated tau CSF concentrations]). Alternatively, it may be possible to employ the recently proposed biomarker-guided A/T/N descriptive classification system¹⁰⁸ in the validation stage.

The predictive accuracy requirements of a screening tool (PPV and NPV) are dependent on COU, in addition to the to the follow-up diagnostic resources, treatments and associated costs. A primary care-based screening tool would not be intended as a “diagnostic”, but would serve as a gatekeeper to the confirmatory diagnostic procedures utilized by specialty physicians. As is the case with all primary care screening tools, the primary concern is with high NPV. This is due to the base rates of the conditions in the general population (which are typically very low) and, therefore, the screening tool serves to preclude the vast majority of patients who are not in need of more invasive and expensive diagnostic procedures. For example, in mammography screening for breast cancer and in depression screening in primary care, PPVs are <30%⁶³, but are counteracted by excellent NPVs (>90%). In primary care settings, it is commonplace to see tools with PPVs <40% (or even <20%) but NPVs >90% (or even >95%)⁶³. The overall diagnostic accuracy of the primary care screening tool must be taken into account alongside the full diagnostic procedures in

specialty clinics where available that include CSF and/or PET biomarkers (which have excellent PPV) for confirmatory diagnostic purposes. Once disease-modifying therapeutics become available (for prevention or treatment of symptomatic patients¹⁰⁹), these primary care screening tools become required as there will be many more patients seeking services. Therefore, the primary care screen will be needed to cost-contain while simultaneously broadly opening up access to specialty diagnostics. The consensus in the working group was that a candidate biomarker for primary care screening should ideally have a PPV of >50%, with an aim for this to be as high as possible. However, given the ethical implications of false negatives, it would be more important to achieve a NPV of 90%, with 95–98% being an ideal target⁶³. On the other hand, a performance with a high PPV and an acceptable NPV may be desirable in a pharmaceutical setting that focuses on identifying as many amyloid-positive patients as possible for either inclusion in to clinical trials or a possible future treatment.

In terms of assay technology, the consensus indicated a strong preference for panel-based assays (i.e. a combination of biomarkers, either developed together or individually) over single biomarkers. Panels may offer wider applications beyond AD diagnosis, allowing more generalized testing for other ND, with each pathological condition having a unique protein signature. Several studies established panels of biomarkers to discriminate between cognitively healthy controls and AD participants and assessed large arrays of differently combined proteins to yield high specificity and sensitivity. Indeed, since there is a definite need for a holistic approach to standardize blood biomarkers for AD diagnosis, it is crucial to understand the link among various individual biomarkers and overcome the outdated approach of examining single candidate biomarkers at a time¹¹⁰. Given that various methods exist to detect biomarkers in blood^{28,110–112}, it is crucial to standardize the technologies generating complex multidimensional data and the whole workflow. General consensus on both protocols and ultrasensitive analytical methods are needed across multicentre studies to set up a standardized panel of biomarkers for AD diagnosis^{110,111}. In this regard, recent technical advances have provided ultrasensitive immunoassay methods that allow detecting and quantifying biomarkers at concentrations that are several-fold lower than those accessible by existing immunoassays. This may result in the development of a new array of blood biomarkers covering the entire *spectrum* of AD molecular pathophysiological mechanisms. Moreover, these technologies may serve as the foundation enabling the diagnosis of AD and other ND earlier than ever before¹¹³.

However, the number of biomarkers in the various types of panels so far developed increases the complexity of manufacture and commercial development, complicates validation and standardization processes, and leads to increased costs. This is a significant concern when considering scalability to meet global primary care needs. Panel-based assays also offer an additional challenge, as there are risks associated with analyzing large numbers of candidate molecules in small numbers of patients, which may result in data over-fitting and misleading results so that a panel seems to perform much better than individual biomarkers, but in fact does not¹¹⁴. This can be partly offset by use of a training set to initially reduce the dimensionality before validation in an independent cohort¹¹⁵. Therefore, while the working group agreed that a multi-marker panel may be needed, it was also agreed that the number of

biomarkers should be restricted as much as possible within the NPV and PPV requirements as this would help in meeting the validation and regulatory hurdles.

Refinement and validation

Following identification of a prioritized candidate biomarker, the biomarker will require refinement and validation. Refinement will involve agreeing on a best-practice protocol, including blood collection procedures, pre-analytical sample handling and procedures for harmonization between different laboratories. Validation steps are outlined below.

Validation considerations and recommendations—The consensus process for a pathway towards validation is shown in Figure 4. Prior to validation for widespread use, a biomarker should already have undergone initial clinical validation after the discovery analyses. Validation of target biomarkers should begin with assessment in a “black-and-white” panel study, in which samples would be obtained from AD patients diagnosed according to the International Working Group-2 (IWG-2) criteria (amyloid PET or CSF AD signature), and also from cognitively normal (CN) controls. Samples would be divided approximately 50:50 CN:AD. This initial study would aim to establish concordance between the novel biomarker and gold standard methods, and would be performed using the original technology employed in the initial development of the biomarker test. Ideally, the study would be conducted by an external investigator blinded to the patient characteristics associated with each sample. Analysis of a black-and-white panel would not test the biomarker within the intended COU, but would instead attempt to independently validate the overall accuracy of the biomarker in a known group design. Notably, screening tests for AD may or may not detect MCI, and the accuracy of detection for MCI is likely to be lower, if even possible⁶³. Therefore, multiple tests may be required if the target population is to include both early AD and MCI.

Following the black-and-white panel, the next step would be to attempt to replicate the initial study results in a more representative sample set, i.e. one reflecting primary care. This second study would again employ the original technology from the developing laboratory while simultaneously beginning to transfer the technology to an existing diagnostic assay platform that is: a) globally available; b) capable of meeting the scalability needs for the numbers of assays needed when reaching primary care providers; and c) developed by an entity with prior experience of undergoing regulatory approval for diagnostics. Technology transfers are likely to be mandatory, as discovery platforms are unlikely to meet the stringent requirements for diagnostic assays as outlined by the Clinical Laboratory and Standards Institute (<http://clsi.org/>). Collection and assay of samples across both the original platform and the intended platform for use downstream in the regulatory pathway creates a built-in “bridge” cohort into the initial trial to test the technology within the intended COU. If strong concordance of results between internal and external laboratories is demonstrated following these studies, data acquired across the original training cohort and the new COU-specific test cohort (using the new technological platform) would be applicable for use in further refinement of the diagnostic algorithm, enabling the case for regulatory approval to be built. This would increase the reliability of the results. Finally, samples would be assayed on the

intended platform for regulatory approval across multiple laboratories in order to establish inter-laboratory reproducibility.

Next, the full regulatory pathway, including all additional trials that might be required, would be established in conjunction with regulatory authorities^{116–118}. An additional validation study within an independent primary care cohort would likely be required for regulatory approval. For US approval, this would need to comply with U.S. Food and Drug Administration (FDA) requirements for registrational studies (prospective data or robustly collected retrospective data). An optional further step would be validation using CSF samples collected from the same patients as included in the original study. Ideally, CSF samples would be collected from both blood-test-negative and blood-test-positive cases in the primary care setting to set the stage for studies examining the utility of the biomarker (or new biomarkers) in detecting amyloid positivity among cognitively normal elders. However, this latter use would require the entire process to begin anew, as this would represent an entirely different COU.

Need for a new cohort?—For the agreed intended use of the putative assay as a primary care screening tool, the ideal cohort to include in pivotal studies would align closely with the actual epidemiology of AD in primary care. The incidence of AD in this setting is ~10%¹, a proportion which includes patients with either subjective or objective cognitive deficits. The ideal cohort should also provide access to PET and/or CSF data for all patients, and have known information for age, sex, *APOE ε4* status and education to ensure that these are taken into account.

Current cohorts with both CSF and blood available are largely based around dementia specialty clinics and, as such, do not meet the needs of the target product profile. Current cohorts are also limited by referral bias – they are enriched for patients with cognitive impairment and so do not echo the disease prevalence in primary care. An alternative to specifically examining primary care cohorts could be to use a yoked design, whereby the setting is a specialist centre with referrals from primary care where, in addition to referred patients, additional patients not meeting the referral criteria could be taken and included to make the cohort more representative. Additionally, it could be possible to work with primary care research networks.

In summary, it may be necessary to recruit new cohorts for the purposes of blood-based biomarker assay development. Alternatively, new large-scale cohorts, such as that created as part of the U.S. “All of Us Research Program”, evolved from the U.S. Precision Medicine Initiative (PMI)¹¹⁹ Cohort Program (available at <https://www.nih.gov/research-training/all-of-us-research-program>), or the UK Biobank¹²⁰, could prove extremely valuable in the development of blood-based biomarkers, as these will represent real-world populations with a wide range of genetic and biomarker data available. Recently, in order to advance the development of the precision medicine paradigm in AD, the international Alzheimer Precision Medicine Initiative (APMI) and its planned pilot-cohort program (APMI-CP25,^{40–43}) have been launched and thematically associated with the “All of Us Research Program”. The mono-centre pilot APMI cohorts, spanning from early asymptomatic preclinical populations to prodromal to late-stage dementia populations, allow for the

standardized recruitment of both cognitively intact individuals at risk for AD and patients with a full *spectrum* of ND to provide an assortment of unique heterogeneous and multidimensional data. Particularly, the “INveStIGATION of AlzHeimer’s PredicTors in Subjective Memory Complainers” (INSIGHT-preAD) study¹²¹ – a French mono-centric academic university-based cohort established at the Institute of Memory and Alzheimer’s disease (*Institut de la Mémoire et de la Maladie d’Alzheimer, IM2A*) at the Pitié-Salpêtrière University Hospital in Paris – is the key cohort of the APMI-CP25,40,41,43. The main goal of the INSIGHT-preAD study is to investigate the earliest preclinical stages of AD and its development, including influencing factors and markers of progression, in cognitively normal Caucasian individuals, recruited from the community in the wider Paris area, France, aged 70 to 85, with subjective complaint of memory dysfunction.

Does the landscape fit the target product profile?

Following the landscape analysis, the 196 candidate biomarkers (Figures 2A and 2B) were further divided into tiers according to additional criteria (outlined in Callout Box 1). Of these 196 candidates, 19 were prioritized for further consideration by the working group (a table of all 19 assays is provided in the Supplementary Information); however, none were deemed to have met the agreed target product profile.

Most of the candidate biomarkers were limited by a lack of validation in independent cohorts (only five biomarker studies identified in the search have been followed up with validation studies) (Figure 5)^{97,98,122–124}. This is regarded as a critical step in development and so is a conspicuous gap. Furthermore, where validation has been performed, it may not have been conducted by an independent group, and may not be based on a model published prior to the study¹⁰⁵. This results in a risk of hypothesizing after results are known (HARKing), which can lead to selective reporting and inflated predictive accuracy¹²⁵.

Validation studies may require additional cohorts or, potentially, they could be validated within subgroups of the original cohort. However, appropriate cohorts for the ideal intended COU do not currently exist; validation between cohorts will therefore be a further hurdle in the development of assays. As previously discussed, the appropriate cohorts (stemming from patients in primary care with CSF and blood samples, as well as PET data) are not available and therefore the target product profile could not be met by any study.

Additional issues present with some biomarkers are PPV and NPV values lower than those specified in the target product profile, as well as small sample sizes (Figure 5). Studies using large and well-established cohorts such as ADNI and the Australian Imaging, Biomarker and Lifestyle Study of Ageing (AIBL) are more likely to meet the initial requirements. As discussed previously, biomarker panels are thought to offer the greatest chances of success, although panels featuring high numbers of markers will be difficult and expensive to develop and validate. In addition, most available technology platforms have a low ceiling on the number of biomarkers that can be contained within the assay, which is usually around five biomarkers. There are many emerging areas of interest (e.g. exosomes) that offer promise, although they are too early for consideration at this time. Additionally, blood-based assays for well-known biomarkers of AD, such as $A\beta_{1-42}/A\beta_{1-40}$, tau, NF-L, and BACE1, may

have potential, and should be monitored closely as more information becomes available and novel technologies emerge to address the hurdles currently facing use of these biomarkers.

The future: blood-based biomarker diagnostics

Blood-based biomarkers have the potential to improve detection and diagnosis for AD by increasing convenience, acceptability and ease of testing, as well as reducing costs. There are a number of key areas in which they are likely to provide most benefit. The first and most important of these is testing for AD in the primary care setting, which will potentially allow identification of patients at the earliest stages of the disease. Ideally, this would be prior to the development of any noticeable cognitive deficits, but this will depend on future developments. The advent of testing in primary care would lead to substantial changes in the current treatment paradigm by allowing patients to be guided towards specialist diagnostics (e.g. CSF or PET scans) and access to care earlier in the course of the disease. This will be particularly important once disease-modifying therapies (DMTs) become available, but may provide benefit to patients and caregivers even at the present time¹²⁶.

Once DMTs become available there will, undoubtedly, be a tremendous surge in numbers of patients in primary care settings seeking potential prescriptions. However, without a convenient primary care screen, confirmatory diagnostics will still be required (i.e. CSF, MRI or PET), which will act as a bottleneck in the diagnosis and treatment process, and will present a considerable burden in terms of costs. Availability of a multi-tiered diagnostic process commencing with a blood test in primary care would dramatically increase access to DMTs, while conveniently exclude from further diagnosis and treatment the vast majority of patients who would not need to undergo these expensive and invasive procedures. An additional potential use of blood-based AD biomarkers will be to provide a cost-effective and rapid test for AD pathology in order to determine the eligibility of patients for recruitment into clinical trials for DMTs^{127–129}. These tests might also have applicability in the monitoring of disease progression and drug effects on AD pathophysiology during a trial¹³⁰. As such, the availability of blood-based biomarker diagnostics may in fact accelerate the development of DMTs.

The development of blood-based biomarkers often stalls in the early stages due to a disconnection between academia, where biomarkers are identified, and industry, who have the resources to take biomarkers further and develop them commercially (Figure 6). This disconnect arises from the different needs of these different sectors and from mutual mistrust developed following a history of poor interactions. The current initiative by the working group and Roche Diagnostics International represents a step in an attempt to improve that process, both efficiently and economically.

Conclusions

The last decade has unquestionably witnessed key progresses in the area of blood-based biomarkers research in AD and other ND. Human blood (plasma) holds the largest source of the proteome and technologies measuring even minor alterations of proteins and peptides are crucial tools to identify protein biomarkers in blood. For instance, mass spectrometry technologies can detect slight alterations of the protein concentrations and

immunohistochemistry can recognize with high accuracy a definite protein in the living system. Compared with CSF markers, a validated blood-based AD biomarker would provide a fast, non-invasive, and cost-effective method of early detection and diagnosis for what is the most common age-related ND worldwide. In addition, venipuncture is a routine, safe procedure that does not pose any harm to the patient. As such – and differently from CSF sampling – the examination of blood biomarkers is accepted and more easily introduced in the clinical environment²⁶. Targeting symptomatic patients in a primary care setting would help patients with early AD access care earlier. Development of biomarkers ideally reflecting all existing molecular pathophysiological mechanisms in the polygenic AD brain at distinct time points of progression will represent the foundation for personalized, tailored, biomarker-guided targeted therapies and constitute a critical step towards the dissemination of precision medicine in AD^{25,40}. Furthermore, the development and implementation of a multi-stage diagnostic approach beginning with a blood test in primary care will increase access to confirmatory diagnostic modalities (such as PET imaging and/or CSF sampling) and provide a clear path for regulatory approval. This tiered approach is expected to increase the access to DMTs, once these will ultimately be available.

Despite much research, many of the candidate blood-based biomarkers exhibit limitations currently preventing their use. Close cooperation and coordination is needed among academic institutions, industry partners, and regulatory bodies to accelerate the development of a blood-based biomarker assay suitable for clinical use. The assembly of novel cohorts may be required to allow the development of a new assay for the agreed ideal intended use of broad population-based screening for AD in patients in primary care.

Applying systems biology to inspect large multidimensional blood-based “omics” data will enable the stratification of patient populations into well-defined subsets of patients sharing a common pathophysiology that, in turn, can be further explored for targeted interventions. The substantial use of systems biology methods to discover and validate diagnostic biomarkers of specific subsets of AD patients will considerably impact the progress of precision medicine in AD and facilitate the use of this paradigm for AD patients towards clinical reality²⁶.

“Omics” sciences – providing a depiction of complex biological systems in a holistic and integrative manner – generate unavoidably large-scale and heterogeneous biomedical data. This led to open the era of “big data” in Biology and Medicine^{25,40,42}, indicating the complexity and challenges presented by the combined analysis of data. These encompass an enormous amount of heterogeneous, disorganized, multidimensional data (from molecular/cellular data, to imaging, clinical, demographic, and environmental data) extensively produced by academic institutions, clinics, and, more recently, mobile devices^{131,132}. The implementation of systems biology to attain novel insights into AD pathophysiology requires the assessment of big and deep multidimensional heterogeneous data. Performing the integrated analysis of big data depends on the accessibility to appropriate tools, both for data storage/management and for their modelling, according to disease pathophysiology. Therefore, effective and sophisticated methods will be essential to systematically screen for novel blood-based biomarkers associated with AD, as well as to gain insights into their spatiotemporal interactions with other biomarker categories. The integration of the results

obtained with different experimental strategies is assumed to provide complementary information on the disease pathophysiology and provide insights into its clinical interpretation. Accordingly, the Big Data Research and Development Initiative (available at <https://obamawhitehouse.archives.gov/blog/2012/03/29/big-data-big-deal>), announced by the previous Obama Administration, is a crucial promoter of the implementation of precision medicine through the integration of big and deep biomedical data. The ability to deal with “big data science”, accompanied by the implementation of an integrative disease modelling¹³³, is an essential aspect of the previously mentioned international APMI and other worldwide initiatives, including the previously mentioned Alzheimer’s Association ISTAART BBB-PIA71 and the recently established Cholinergic System Working Group (CSWG)¹³⁴.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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- *In Vitro* Procedure for Diagnosis and Early Diagnosis of Neurodegenerative Diseases Patent Number: 8298784
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- *In Vitro* Multiparameter Determination Method for The Diagnosis and Early Diagnosis of Neurodegenerative Disorders Publication Number: 20100062463
- *In Vitro* Method for The Diagnosis and Early Diagnosis of Neurodegenerative Disorders Publication Number: 20100035286

- *In Vitro* Procedure for Diagnosis and Early Diagnosis of Neurodegenerative Diseases Publication Number: 20090263822
- *In Vitro* Method for The Diagnosis of Neurodegenerative Diseases Patent Number: 7547553
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Callout Box 1**Criteria for evaluation of identified biomarkers**

- Tier 1 – validated biomarkers
 - Markers with strong performance data that have successfully completed initial validation studies in subsequent reports or in a multi-centre setting with more than 120 patients.
- Tier 2 – high-performing biomarkers
 - Markers with good performance data that have not completed validation studies. Markers in this category are further divided into subgroups depending on the intended use and number of patients.
 - 2A: Specificity **and** sensitivity >80%, or specificity **or** sensitivity >85%^{*}, with an intended use of early detection, prediction or differential diagnosis and **greater than** 120 subjects
 - 2B: Specificity **and** sensitivity >80%, or specificity **or** sensitivity >85%^{*}, with an intended use of early detection, prediction or differential diagnosis and **less than** 120 subjects
 - 2C: Specificity **and** sensitivity >80%, or specificity **or** sensitivity >85%^{*}, with an intended use of diagnosis
 - 2D: Same criteria as Tier 2A–C, with markers having applications in both blood and CSF.
- Tier 3 – promising candidate biomarkers
 - Innovative markers with promising performance data but have not yet produced higher performance results in clinical settings.

^{*} Along with an adequate, non-zero score on the other parameter

Callout Box 2**Checklist for developing a blood-based biomarker**

- Define context of use and setting
 - Define attributes of biomarker (how it enters the blood, concentrations and diurnal changes)
 - Develop detection method
 - Validate against gold standard (IWG-2 criteria)
 - Replicate in target-setting
 - Refine to reach target predictive values
 - Demonstration of technical performance
 - Validate against large, independent cohorts
- Validated biomarkers should have a study showing a difference between AD and healthy controls and diagnostic utility in subjects with MCI

Glossary

Technical definitions

α -Synuclein

α -Synuclein is a neuronal protein involved in regulating the synaptic vesicles pool size, trafficking, membrane dynamics, and neurotransmitters release. Intracellular aggregates of α -synuclein are typical of certain neurodegenerative diseases (α -synucleinopathies) including Parkinson's disease, dementia with Lewy bodies, and multiple system atrophy. α -Synuclein aggregates are also found in other neurodegenerative diseases, including AD.

ADNI

The Alzheimer's Disease Neuroimaging Initiative (ADNI) is an ongoing, longitudinal, multicentre study, initiated in the U.S.A. in 2004, designed to develop and validate clinical, imaging, biochemical, and genetic, biomarkers for the early detection and tracking of AD.

AIBL

The Australian Imaging, Biomarker and Lifestyle Study of Ageing (AIBL) is an ongoing, longitudinal study, initiated in Australia in 2006, exploring biomarkers, cognitive characteristics, as well as health and lifestyle factors associated with AD.

Amyloid precursor protein (APP)

APP is a type I membrane protein abundantly expressed in the brain and implicated in neural growth and maturation. APP can be processed via the non-amyloidogenic pathway, where cleavage by α - and γ -secretases generate a secreted form of APP (sAPP α) and C-terminal fragments (CTF 83, p3 and AICD50). Alternatively, APP can be processed via the amyloidogenic pathway, where cleavage by β - and γ -secretases generate a secreted form of APP (sAPP β), C-terminal fragments (CTF 99 and CTF 89) and A β peptides.

Amyloid beta (A β)

A β peptides are produced in variable length, including the 42-amino acid-long and the 40-amino acid-long amyloid beta peptides (A β ₁₋₄₀ and A β ₁₋₄₂), from the proteolytic cleavage of the amyloid precursor protein. A β ₁₋₄₂ exhibits a substantial propensity to form aggregates, including extracellular amyloid plaques, which are a pathological hallmark of AD.

A/T/N

The A/T/N scheme is a classification system that uses three binary biomarker categories reflecting AD pathophysiology. 'A' refers to biomarkers of amyloid beta pathology (CSF A β ₁₋₄₂ or amyloid PET), 'T' refers to biomarkers of tau pathology (CSF hyperphosphorylated tau or tau PET), and 'N' refers to biomarkers of neurodegeneration or neuronal injury (CSF total tau, ¹⁸F-fluorodeoxyglucose-PET (¹⁸F-FDG-PET), structural MRI).108

β -site amyloid precursor protein cleaving enzyme 1 (BACE1)

BACE1 is an endoprotease responsible for the first proteolytic cleaving step in the generation of A β peptides from the amyloid precursor protein.

BioFINDER

The Biomarkers For Identifying Neurodegenerative Disorders Early and Reliably (BioFINDER) study is an ongoing longitudinal study, initiated in Sweden in 2010, investigating the early diagnosis and underlying pathologies of AD and Parkinson's disease based on neuroimaging, cerebrospinal fluid, blood-based, clinical and neuropsychological markers.

Cerebrospinal fluid (CSF)

CSF is the body fluid within the central canal of the spinal cord and the four ventricles of the brain.

Context of use (COU)

A COU is a statement that describes the manner and purpose of use for the biomarker in drug development. The supporting data and analyses submitted with the biomarker qualification determine the acceptability of the qualified COU.

IWG-2

The IWG-2 criteria are research diagnostic criteria for AD, developed by the International Working Group (IWG) for the diagnosis of AD.²⁹

Mild cognitive impairment (MCI)

MCI causes a slight but noticeable and measurable decline in cognitive abilities, including memory and thinking skills. An individual with MCI is at an increased risk of developing AD or another form of dementia.

Negative predictive value (NPV)

The NPV is the probability a patient does not have the disease when the test result is negative.

Neurofilament light (NF-L)

NF-L chain protein is a component of neurofilaments, which are structural constituents of the neuronal cytoskeleton. It is highly expressed in large-calibre myelinated axons and is released into the brain interstitial fluid, following axonal injury.

Plasma

Plasma is the liquid constituent of blood obtained after removal of blood cells.

Positive predictive value (PPV)

The PPV is the probability a patient has the disease when the test result is positive.

Receiver operating characteristic (ROC)

The ROC curve is a plot of sensitivity versus 1-specificity for the different possible cut-off points of a diagnostic test. Accuracy of the diagnostic test is based on the area under the ROC curve; the closer the area under the ROC curve is to 1, the better the test.

Sensitivity

Diagnostic sensitivity is the probability a test result is positive when the disease is present.

Serum

Serum is the clear liquid constituent of blood characterized by the absence of the coagulation factors and blood cells

Specificity

Diagnostic specificity is the probability a test result is negative when the disease is absent.

Subjective cognitive decline (SCD)

SCD is a self-reported decline in cognition, undetected by standard neuropsychological tests.

Tau

Tau is a microtubule-associated protein that modulates the stability of axonal microtubules. In AD and other neurodegenerative diseases (tauopathies), aggregates of abnormally phosphorylated (hyperphosphorylated) tau proteins result in intraneuronal neurofibrillary tangles.

TDP-43

Transactive response DNA-binding protein 43 (TDP-43) binds both DNA and RNA and helps regulate mechanisms of RNA processing such as splicing, trafficking, and microRNA production. TDP-43 cytoplasmic inclusions are associated with definite neurodegenerative diseases, including frontotemporal dementia, amyotrophic lateral sclerosis, and AD.

Conceptual definitions

Big Data

Big Data is a repository of large amounts of data sets generated by data mining tools. Big Data includes information obtained through systems theory- and knowledge-based approaches, and clinical records.

Data science

Data science is an interdisciplinary field about processes and systems to extract knowledge from data in different forms – either structured or unstructured – which is a continuation of some of the data analysis fields including statistics, artificial intelligence, machine learning, data mining, and predictive analytics.

Homeostasis

Homeostasis is a spontaneous tendency towards a condition of a dynamic equilibrium based on a continuous counterbalance between regulatory-defense mechanisms and disrupting stress-induced signals. Homeostasis is common to any biological system. Homeostatic signalling is hierarchically organized from subcellular to cellular level, across organs, and, finally, systems. Homeostasis is essential for protecting all core biosynthetic processes necessary to optimal functioning and survival.

Integrative disease modelling

Integrative disease modelling is a multidisciplinary approach to standardize, manage, integrate, and interpret multiple sources of structured and unstructured quantitative and qualitative data across biological scales using computational models that assist decision making for translation of patient-specific molecular mechanisms into tailored clinical applications.

Network

A network is a set of recurring motifs; each motif is a pathway; each pathway, in turn, carries out specific dynamical functions and can be modulated, i.e. up-down regulated, either upstream or downstream or both. There is a large spectrum of biological cross-talks between pathways and networks inside a level of a given system and among systems.

“Omics”

“Omics” are high-throughput screening tools aimed at fully collecting, characterizing and quantifying pools of biological molecules (DNA sequences, transcripts, miRNAs, proteins/peptides, metabolites/lipids) that translate into the structure, function, and dynamics of an organism and/or whole organisms.

“One-size-fits-all” approach

The “one-size-fits-all” approach is the traditional approach used for the development of early detection, intervention, and prevention options, where biomarker candidates are being validated against the plethora of heterogeneous clinical operationalized syndromes, rather than against genetically (risk profile) and biologically (i.e., based on molecular mechanisms and cellular pathways) determined entities.

Precision medicine

Precision medicine is a translational science paradigm related to both health and disease. It is a biomarker-guided targeted medicine on systems-levels taking into account methodological advancements and discoveries of the comprehensive pathophysiological profiles of complex polygenic, multi-factorial neurodegenerative diseases (proteinopathies of the brain). Precision medicine aims to optimize the effectiveness of disease prevention and therapy, by considering (customizing) an individual’s specific “biological make-up” (e.g. genetic, biochemical, phenotypic, lifestyle, and psychosocial characteristics) for targeted interventions through P4M (Predictive, Preventive, Personalized, and Participatory) implementation.

Systems biology

Systems biology is an evolving hypothesis-free, exploratory, holistic (non-reductionistic), global, integrative, and interdisciplinary paradigm using advances in multimodal high-throughput technological platforms that enable the examination of networks of biological pathways where elevated amounts of structurally and functionally different molecules are simultaneously explored over time at a system level (i.e., at the level of molecules and subcellular compartments, cells, group of cells, tissues, organs, apparatuses, or even whole organisms). According to systems biology, organisms are made of systems which are entities consisting in hierarchically self-organized levels with increasing structural complexity resulting in different emerging properties.

Systems theory

Systems theory is a translational research theory of the precision medicine paradigm. It is an interdisciplinary conceptual framework allowing for the conceptualization of novel/original models to extract and explicate all systems levels and different spatiotemporal data types of complex polygenic diseases.

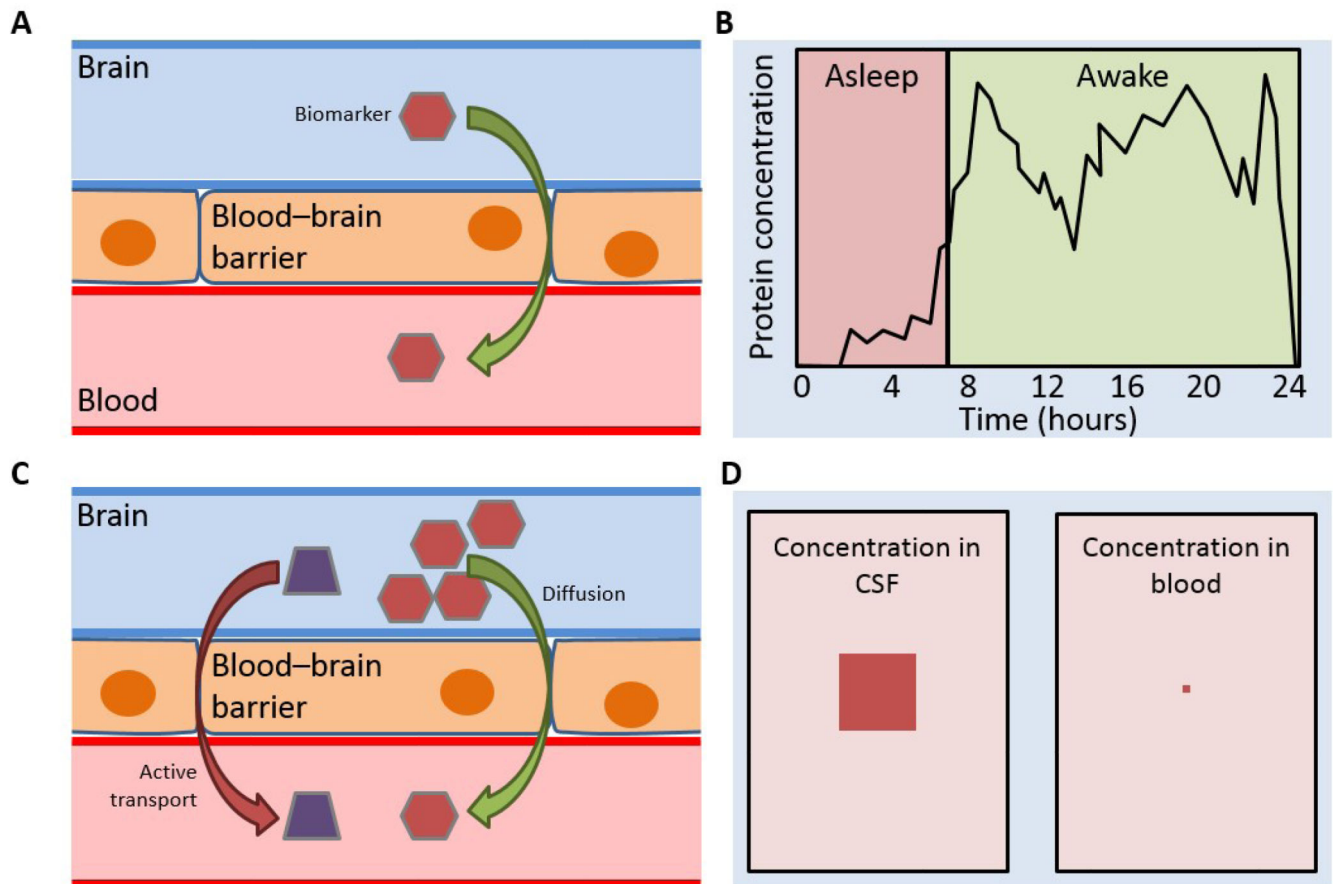


Figure 1. Challenges in developing a blood-based biomarker of CNS disease.

There are several considerations when finding a biomarker for a CNS disease, particularly if the biomarker is to be blood-based.

Blood–brain barrier: biomarker must be able to cross the blood–brain barrier to allow detection (**A**). **Diurnal differences in CSF and blood:** diurnal differences in protein concentrations exist in both CSF and blood. Biomarker levels may not peak at the same time in CSF as in blood, and a biomarker assay will either require sampling at the peak concentration, or sufficient sensitivity to detect the biomarker throughout the day (**B**).

Active/passive transport: biomarkers may originate in either CSF or blood. Those originating in CSF may enter the blood via active or passive transport, and understanding the exact nature of the derivation from CSF will be essential to develop an assay (**C**). **Concentration differences:** biomarker levels are not always similar in CSF and blood. For example, A β concentrations are 10-fold lower in plasma than in CSF (35 pg/ml vs. 350 pg/ml) (**D**).

Abbreviations: A β =amyloid beta; CNS=central nervous system; CSF=cerebrospinal fluid.

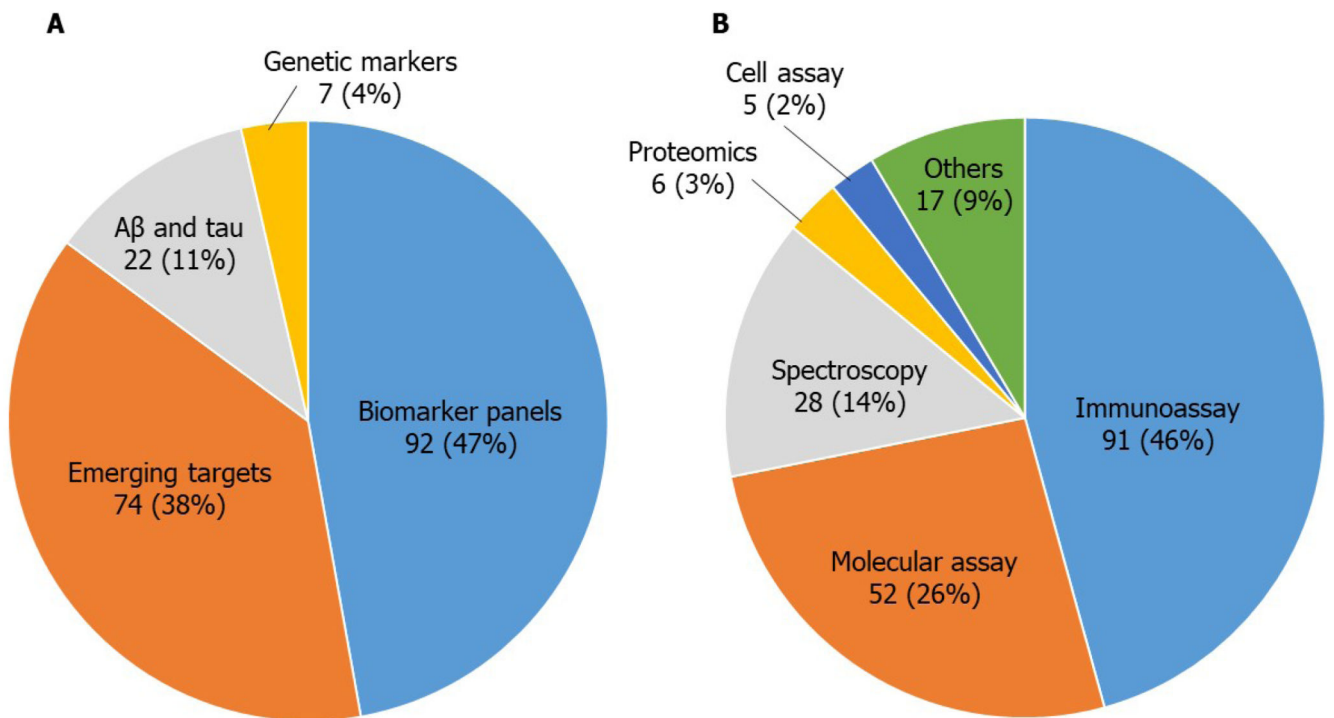


Figure 2. Candidate blood-based biomarkers identified in the landscape analysis. Technology (**A**) and platform (**B**) categories for the 196 high-quality studies identified in the landscape analysis.

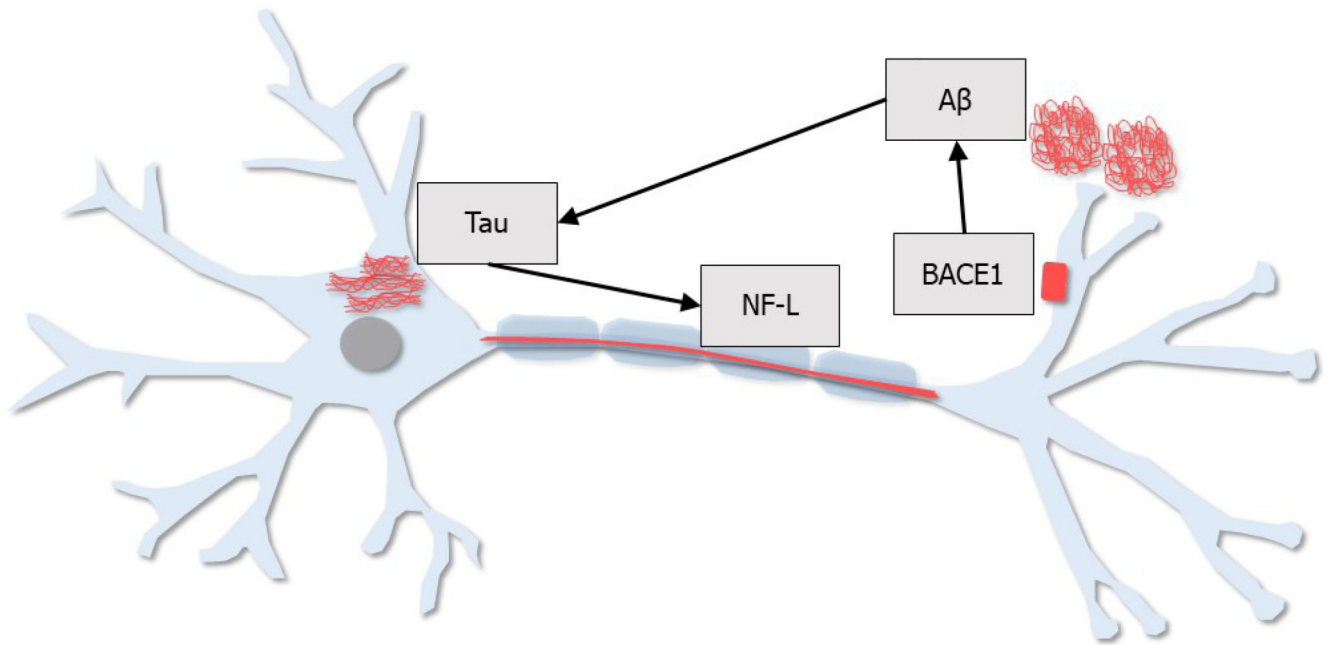


Figure 3. Schematic representation of promising blood-based biomarker candidates

Four promising blood-based biomarker candidates are represented in this schematic:

A β : A β peptides, in particular A β_{1-42} , are implicated in AD pathogenesis; however, it is the ratio of A β_{1-42} /A β_{1-40} that appears to be the most promising A β -related biomarker in the blood.

BACE1: The first step in the generation of A β peptides is the cleavage of APP by the β -secretase, BACE1. Measurement of BACE1 activity in the blood may be useful for predicting progression from MCI to AD dementia.

Tau: Phosphorylated tau protein is a major component of intraneuronal neurofibrillary tangles, which are often present in AD. The abnormal phosphorylation of tau is thought to be driven by A β peptides. Tau levels in the blood may be useful as a predictor of future cognitive decline.

NF-L: NF-L is an axonal protein, released into the brain interstitial fluid following neuronal/axonal injury. NF-L levels in the blood are elevated in AD and other neurodegenerative diseases, therefore, blood-based NF-L could be useful as a biomarker of neurodegeneration.

Abbreviations: A β =amyloid beta; AD=Alzheimer's disease; APP=amyloid precursor protein; BACE1= β -site amyloid precursor protein cleaving enzyme 1; MCI=mild cognitive impairment; NF-L=neurofilament light.

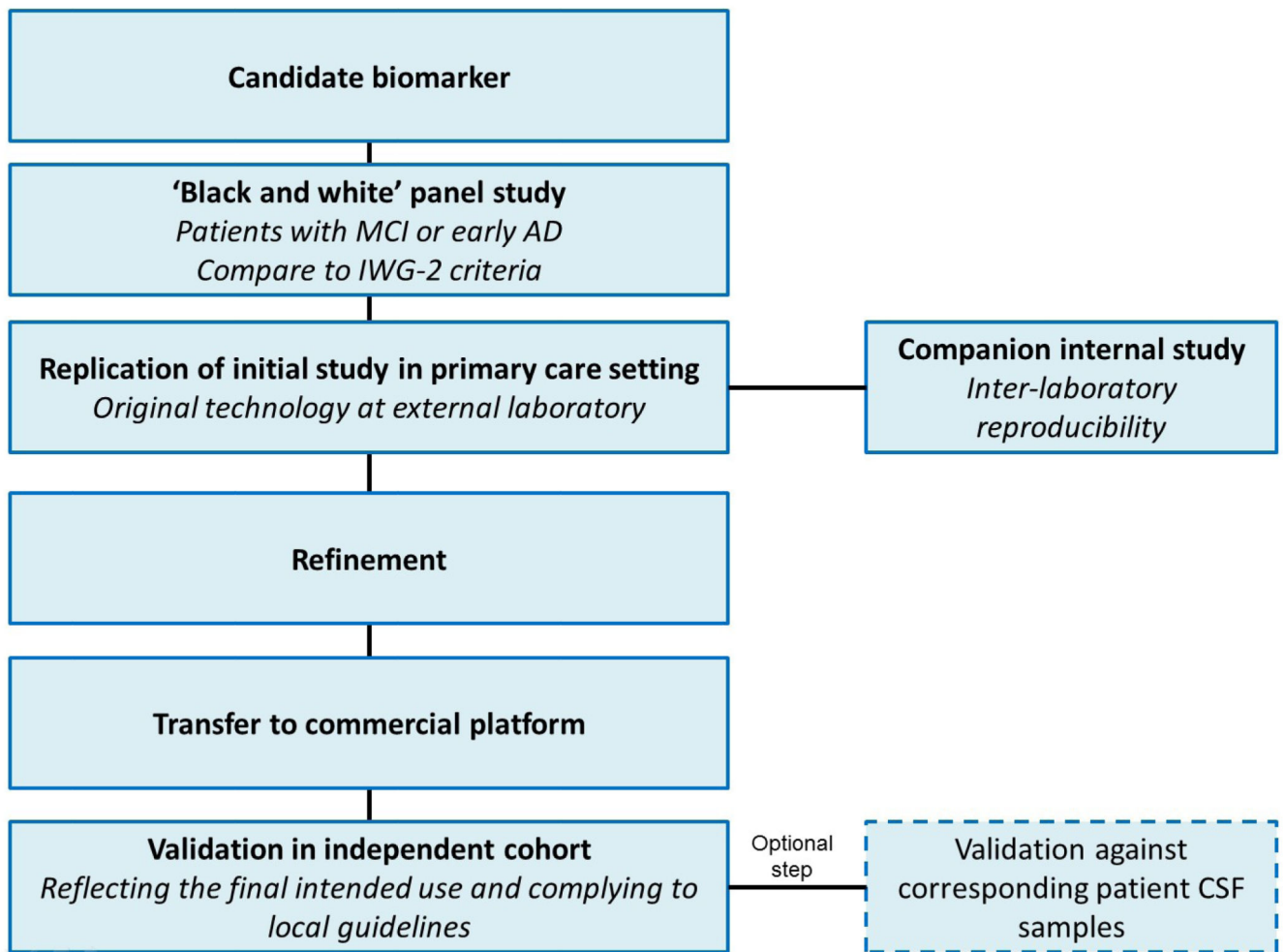


Figure 4. Idealized validation process for blood-based biomarkers.

Abbreviations: AD=Alzheimer's disease; CSF=cerebrospinal fluid; MCI=mild cognitive impairment.

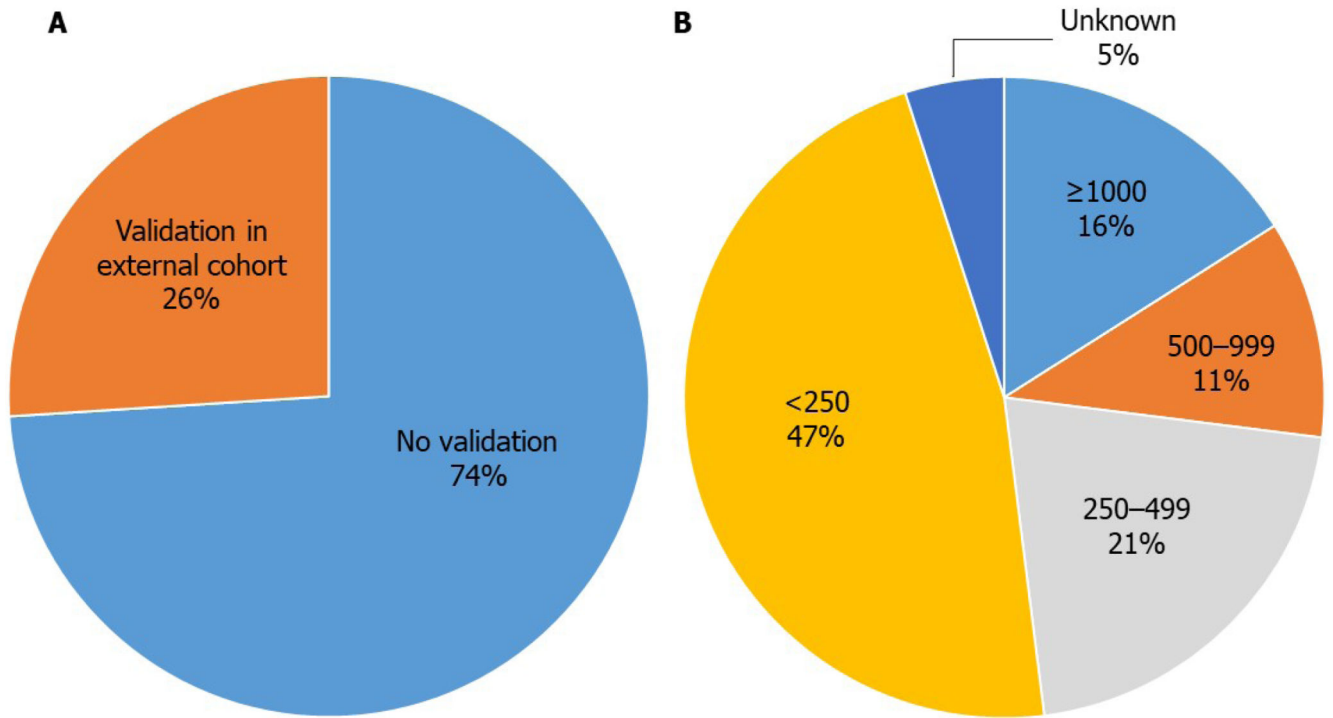


Figure 5. Validation status and patient numbers of the 19 prioritized biomarkers.

Few of the current blood-based biomarker assays in development have undergone validation in an external cohort. Of the 26% of assays that had some degree of external validation, only 2 (10.5%) were in large cohorts such as ADNI or AIBL (**A**). Studies are often underpowered, with small patient numbers (**B**).

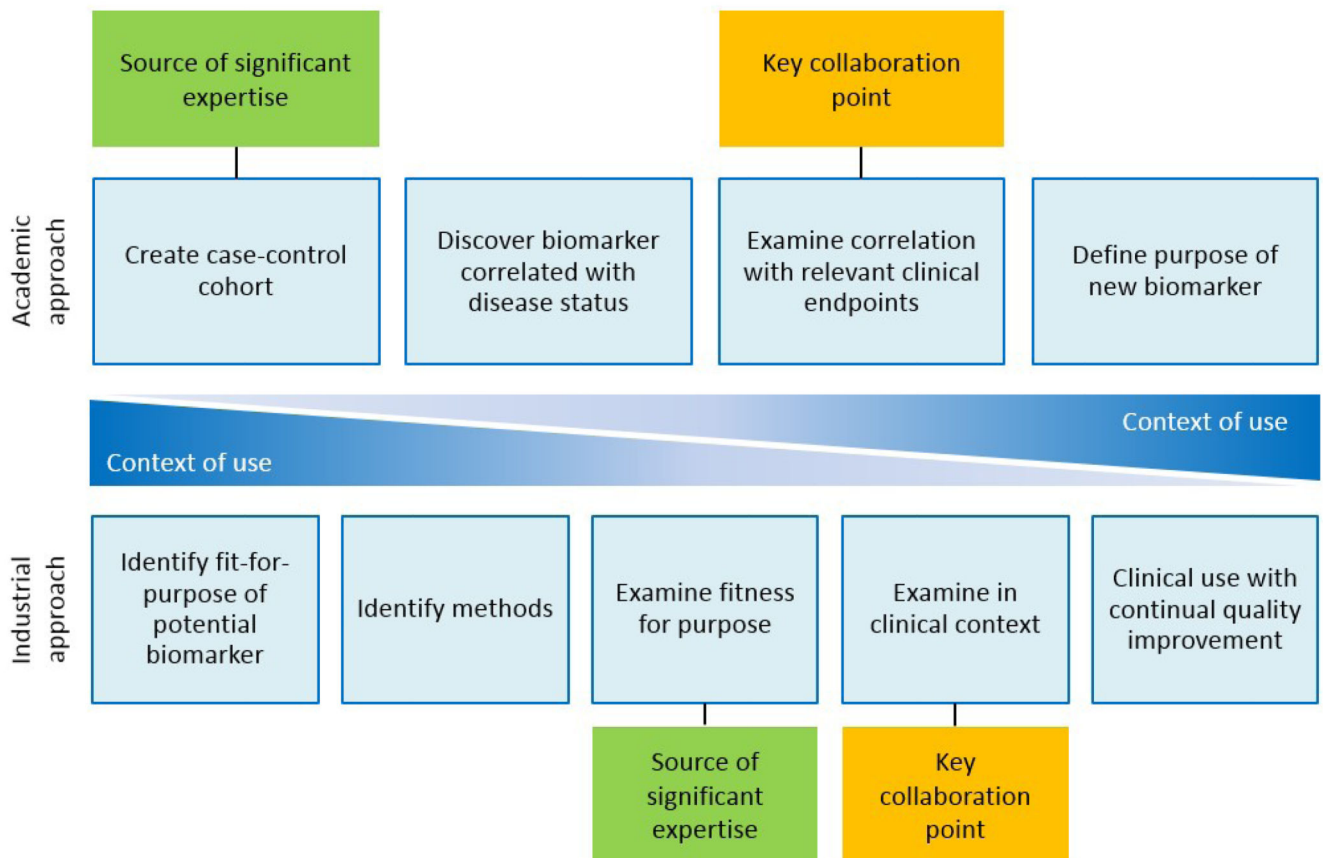


Figure 6. Potential collaboration points between academia and industry.

Academic and industrial approaches to biomarker development are inherently different, but could be extremely useful if combined. Close collaboration between industry and academia would allow for sharing of expertise in product testing, access to cohorts and clinical data, as well as allow for sharing ideas and theories with regards to clinical endpoints and context. By merging the two approaches, a method where the context of use is the primary focus throughout the process can be established. This would allow for synergistic development of a new biomarker between academics and industrial partners, sharing a wealth of experience.

Table 1
Pre-analytical factors influencing results of blood-based biomarker measurements.

Patient-related factors	Blood collection	Blood processing	Sample storage
Demographics (age, sex, weight, ethnicity)	Needle gauge	Time to storage	Storage temperature
Diet, including supplements	Needle composition	Centrifugation (presence and type of separator, temperature, number of rounds [single or double])	Storage volume
Health status	Withdrawal site	Addition of protease inhibitor	Number of freeze-thaw cycles
Medication	Collection tube characteristics	Use of denaturation step or protein extraction	Duration
Drug/alcohol use	Anticoagulant use	Use of plasma or serum	
Medical conditions	Time of collection		
Exercise			
Posture/bed rest			