

ProteonanoTM: a novel deep proteomics platform with picogram sensitivity and its application in Alzheimer's disease Dan Liu^{1, 2, 3}, Yonghao Zhang⁴, Xiehua Ouyang⁴, Shanshan Lv⁴, Yanting Meng⁴, Libing Wang⁴, Zhiquan Cao⁴, Yi Wang⁴, Hao Wu^{4#}, Yan Zeng^{1,2,3#}

ABSTRACT

Introduction: Alzheimer's disease (AD), the predominant form of dementia, manifests as a protracted progression exceeding a decade. It is documented that concomitant cerebral damage accrues prior to the manifestation of overt behavioral symptoms. Timely disease detection is paramount for impeding progression and augmenting patient quality of life. However, extant biomarkers, notably amyloid β (Aβ40, Aβ42), demonstrate suboptimal performance in identifying individuals at early stage of disease development. Thus, it is critical to identify novel biomarkers that distinguish patients at distinct stages of disease development. To meet this demand, we developed a deep, untargeted plasma proteome profiling technology (ProteonanoTM platform) to facilitate early detection of AD.

Methods: Proteonano[™] technology was developed as an affinity-selective mass spectrometry platform, including usage of nanoparticle-based affinity protein binders (nanobinders) to enrich low abundance proteins and employment of an automated pre-treatment workstation for parallel sample preparation. Patients were serially enrolled under approved ethical review. Plasma samples were collected at time of enrollment, and participants were stratified as normal (N), mild cognitive impairment (MCI), and dementia (D). Plasma samples were processed through the Proteonano[™] pipeline and analyzed by a ThermoFisher Orbitrap Astral mass spectrometer at data independent acquisition mode. Raw data were analyzed by using DIA-NN, normalized, further processed by using a customized biostastic and bioinformatic pipeline.

Results: Plasma samples from 206 serially enrolled participants (N=142, MCI=35, D=29) were used for untargeted proteomic analysis by using the ProteonanoTM platform. 4347 protein groups were identified, with 2344 ± 37 (AVG ± SE) protein groups identified in each sample. 2704 of the protein groups were mapped to human plasma protein project published plasma protein database. Concentrations of these proteins spanned nine orders of magnitude, and the lowest abundant protein had a reported plasma concentration of 1.6 pg/mL. Differential protein expression analyses showed 64, 91, and 159 proteins had different abundance between MCI and N, D and MCI, and D and N groups, respectively. The most upregulated protein in MCI group relative to N group was matrix Gla protein (MGP) and most downregulated protein was non-erythrocytic β spectrin (SPTBN1). Eight feature selection methods, including least absolute shrinkage and selection operator (LASSO), and random forest (RF), were employed, and top features selected in each method were subjected to Akaike information criteria (AIC) based model selection. Identified features were then combined for another round of selection. Differentiating powers of these models were assessed by receiver operating curve (ROC) and precision and recall (PR) methods. The best model differentiating MCI and N groups contains nine proteins, with a ROC-AUC of 0.92 (95 % confidence interval: 0.89-0.98), and PR-AUC of 0.84. This model is superior to the discriminative power of A β 40, with ROC-AUC=0.80. Similarly, excellent models differentiating D and N (8 proteins, ROC-AUC=0.99, PR-AUC=0.96) and D and MCI groups (9 proteins, ROC-AUC=0.98, PR-AUC=0.97) were identified. These results indicate a nanobinder assisted untargeted proteomics approach can effectively identify protein features differentiating patients at different cognitive states, which will eventually lead to improved early detection of patients with elevated risk for AD and dementia.

Novel aspect: Proteonano[™] platform is a powerful biomarker discovery tool, which enabled identifying a highly discriminative protein panel for AD early detection.





(D) Distribution of protein abundance of identified protein groups in all samples. Protein concentrations were referenced to the HPPP database. (E) Distribution of protein abundance in each of the patient within the cohort prior to data normalization.

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> Figure 4. Candidate protein GO term enrichment analysis. GO term analysis identified multiple pathway differences between MCI and N patients.





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indicate more reliable predictions.



Relative abundance of top 4 proteins with highest AUC values. (C) ROC curves of top 4 proteins with highest AUC values.

