

BIOMARKERS (NON-NEUROIMAGING)

ProtBAG: eleven organ-specific proteome-based biological age using CSF proteomics

Sarah Ko¹ | Hui Cao² | Mehrshad Saadatinia¹ | Gao Wang³ | Elisa Konofagou³ | Donald Edmondson³ | Wenjia Bai⁴ | Arthur W. Toga⁵ | Christiane Reitz⁶ | Ye Ella Tian⁷ | Andrew Zalesky⁷ | Christos Davatzikos⁸ | Junhao Wen^{1,9}

¹Laboratory of AI and Biomedical Science (LABS), Columbia University, New York, NY, USA

²LABS, Columbia University, New York, NY, USA

³Columbia University, New York, NY, USA

⁴ICL, London, London, UK

⁵Laboratory of Neuro Imaging, Stevens Neuroimaging and Informatics Institute, Keck School of Medicine, University of Southern California, Los Angeles, CA, USA

⁶Columbia University, The Taub Institute for Research on Alzheimer's Disease and the Aging Brain, The Gertrude H. Sergievsky Center and Departments of Neurology and Epidemiology, New York, NY, USA

⁷Melbourne Neuropsychiatry Centre, Department of Psychiatry, Melbourne Medical School, The University of Melbourne, Melbourne, VIC, Australia

⁸Center for AI and Data Science for Integrated Diagnostics, University of Pennsylvania, Philadelphia, PA, USA

⁹New York Genome Center, New York, NY, USA

Correspondence

Junhao Wen, Laboratory of AI and Biomedical Science (LABS), Columbia University, New York, NY, USA.

Email: jw4745@cumc.columbia.edu

Abstract

Background: Recent research^{1,2} has generated increasing interest in modeling human aging and disease within a multi-organ framework. Plasma proteomics³ has emerged as a widely used approach for predicting individual chronological age, resulting in the proteome-based biological age gap (ProtBAG). Here, we used CSF proteomics from the ADNI study to derive 11 organ-specific ProtBAGs using 2 machine learning (ML) methods.

Method: CSF proteomics was generated using the SomaScan 7k platform in ADNI, which included 7,008 protein levels from 736 participants (mean age: 73.3 ± 7.4 years; 57% women). Missing proteomics values were imputed using AutoComplete⁴. Organ-enriched proteins for 11 organ systems were determined by at least four-fold higher mRNA levels in the tissue of interest than in other organ tissues. These organ-enriched proteins were then fit to a linear support vector regression (SVR) and LASSO regression model. Nested random holdout cross-validation (50 repetitions) was implemented; mean absolute error (MAE) and Pearson's r were used to evaluate model performance.

Result: For proteomics imputation, we chose the imputed results with the copy-mask amount of 0.3 (2% missing rate in data), which led to the best model performance ($r^2=0.54$). Overall, LASSO and Linear SVR achieved comparable MAE values with only slight differences between the two models. The brain and hepatic showed the lowest MAE values (Linear SVR: 4.56 and 4.52 for the brain and hepatic ProtBAGs; LASSO 4.55 and 4.52 for the brain and hepatic ProtBAGs) (Figure 1). Across the 11 organ systems, MAE values from our analyses, ranging from 4.5 to 6, were in line with previous literature using brain imaging². In addition, the brain and hepatic showed the highest Pearson's r values (Linear SVR: 0.62 and 0.64 for the brain and hepatic ProtBAGs; LASSO 0.63 and 0.65 for the brain and hepatic ProtBAGs) (Figure 2).

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Other organ systems, such as the endocrine, female reproductive system, and male reproductive system, showed relatively inferior model performance.

Conclusion: This study leverages CSF proteomics data from ADNI to accurately develop 11 organ-specific ProtBAGs, enriching the organ aging clock framework established in previous literature using plasma proteomics. Future research will investigate the relationship between these ProtBAGs, cognition, and AD progression.

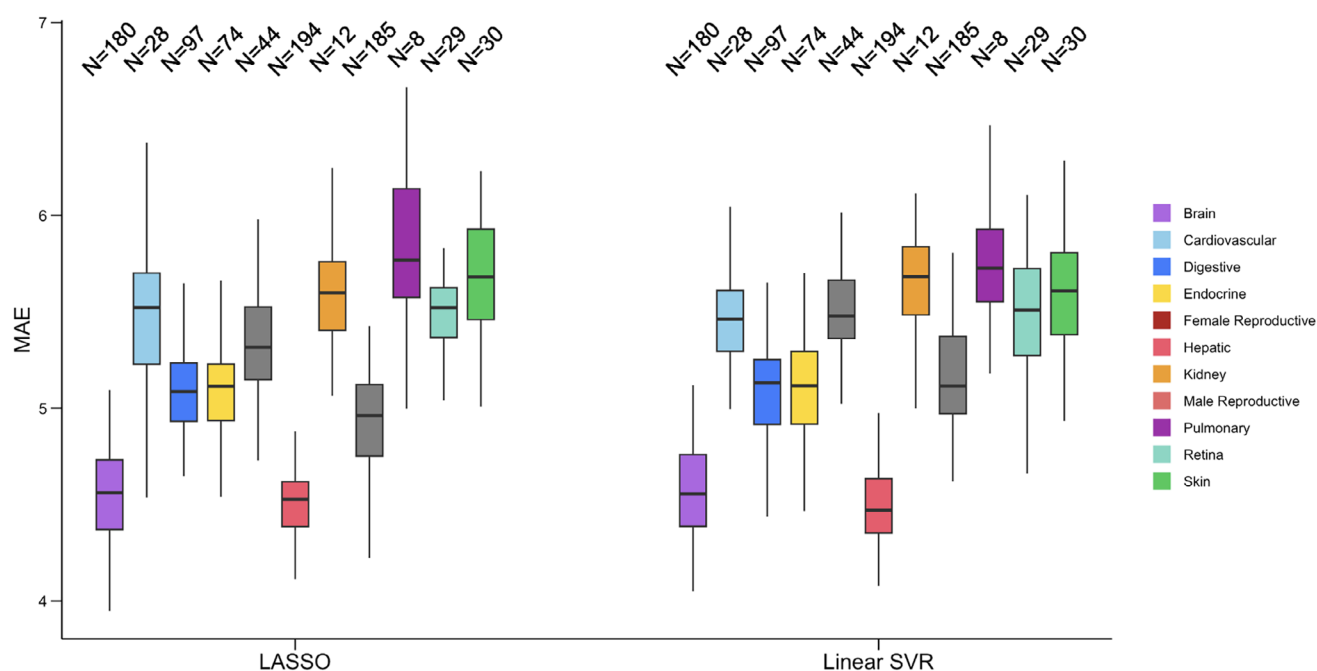


Figure 1. Mean absolute error (MAE) for biological age prediction across 11 organ systems using LASSO and linear SVR. The number of organ-enriched proteins (N) for each organ is displayed. Results are reported without correcting the age bias towards the regression mean.

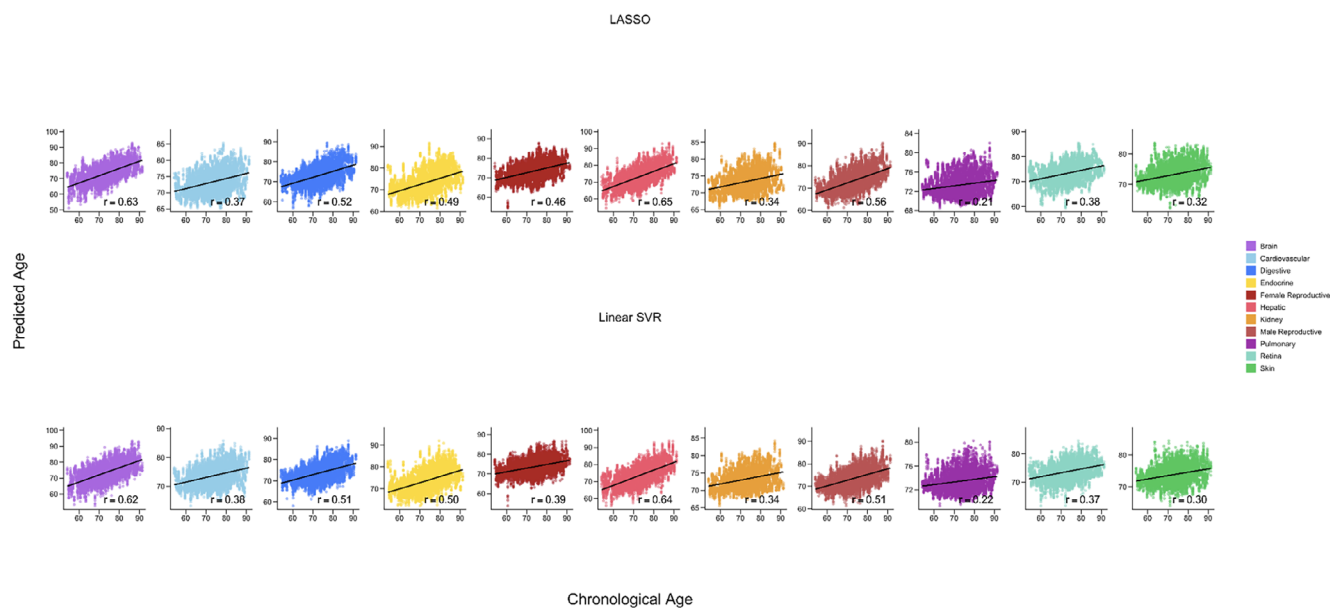


Figure 2. Pearson's r between ML-predicted biological age and chronological age across 11 organ systems using LASSO and linear SVR. Results are reported without correcting the age bias towards the regression mean.