

Predicting multi-omics from genotypes with OmicsPred

April 3 2023, by Justin Jackson

a Platforms with Genetic Scores

Somalogic

Proteomics (plasma)

2,384 protein genetic scores, validated on the FENLAND, MEC and JHS cohorts

[Learn more](#)

Olink

Proteomics (plasma)

308 protein genetic scores, validated on the NSPHG and ORCADES cohorts

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Metabolon

Metabolomics (plasma)

726 metabolite genetic scores, validated on a withheld subset of INTERVAL and ORCADES cohort.

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Nightingale

Metabolomics (serum)

Genetic scores for 141 metabolic traits, validated on UK Biobank, ORCADES and VIKING cohorts

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RNASeq

Transcriptomics (whole blood)

13,668 gene expression genetic scores, validated on a withheld subset of INTERVAL.

[Learn more](#)

c Summary

Number of proteins: 2,384

Training cohort: INTERVAL

Training sample size: 3,175

External validation in cohort 1: FENLAND (European; 2,129 proteins; 8,832 participants)

External validation in cohort 2: MEC (2,047 proteins) - Chinese (CN; N=645), Indian (IN; N=564) and Malay (MA; N=563)

External validation in cohort 3: JHS (African American; 820 proteins; 1,852 participants)

Evaluation Metric: Spearman correlation coefficient (Rho), Variance explained (R²)

b Explore traits and their genetic scores

Find your traits quickly

OmicsPred ID	Somalogic ID	UniProt ID	Gene	Protein	R2AF	Interval R2	Interval Rho	FENLAND R2	FENLAND Rho	FENLAND R2	FENLAND Rho
OP02000001	OLEC10A.1187.11.3	G5GQZ9	OLEC10A	C-type lectin domain family 12 member A	134	0.766	0.864	0.127	0.841	0	0
OP02000002	SGLEC1.3007.7.2	O9Y358	SGLEC9	Sialic acid-binding α -like lectin 9	72	0.746	0.823	0.71	0.809	0.014	0.014
OP02000003	FCGR2A.3309.2.2	P12318	FCGR2A	Low affinity immunoglobulin gamma Fc region receptor 2a	276	0.743	0.845	0.73	0.844	0.04	0.04
OP02000004	ABO.3033.52.3	P16442	ABO	Histo-blood group ABO system transferase	202	0.743	0.837	0.697	0.825	0.03	0.03
OP02000005	FCGR2B.3310.62.1	P31994	FCGR2B	Low affinity immunoglobulin gamma Fc region receptor 2b	127	0.727	0.876	0.128	0.872	0.04	0.04
OP02000006	LLPL8.7015.8.3	O73023	LLPL8	Leukocyte immunoglobulin-like receptor subfamily B member 8	139	0.725	0.847	0.121	0.859	0.04	0.04
OP02000007	PLRA.9427.8.3	GRUKJ1	PLRA	Paired immunoglobulin-like type 2 receptor alpha isoform PDZ-deltaTM	425	0.709	0.808	0.659	0.792	0.04	0.04
OP02000008	ICAM1.4240.13.3	P05062	ICAM1	Intercellular adhesion molecule 1	165	0.699	0.820	0.712	0.841	0.010	0.010
OP02000009	PLRA.10915.195.3	GRUKJ1	PLRA	Paired immunoglobulin-like type 2 receptor alpha isoform PDZ- δ 1+ δ 2	408	0.694	0.791	0.669	0.794	0.008	0.008
OP02000010	SUPPL.2367.7.3	O98028	SUPPL	Tyrosin-related protein	180	0.691	0.857	0.128	0.873	0.161	0.161
OP02000011	MST1.4407.16.1	P28927	MST1	Heterozygote growth factor-like protein	493	0.683	0.839	0.671	0.867	0.046	0.046

d Download

[Download results](#)

[Download model files](#)

Key features of the OmicsPred portal for accessing genetic scores of multi-omic traits. a, Organization of genetic scores on the portal. b, Example of how biomolecular traits and their genetic-score-related information can be explored. c, Example of how summary statistics of training and validation cohorts are presented. d, Example of how validation results and genetic-score models can be downloaded. e, Example of how validation results and trait-related information can be visualized. Credit: *Nature* (2023). DOI: 10.1038/s41586-023-05844-9

Work by an international research team led by Yu Xu and Michael Inouye at the Department of Public Health and Primary Care, University of Cambridge, has resulted in a unique resource for predicting multi-omics data directly from genotypes. The details of how the team developed the OmicsPred resource are in an article, "An atlas of genetic scores to predict multi-omic traits," published in the journal *Nature*. A Research Briefing on the study is published in the same journal issue.

There is a vision of the future that includes a device like the medical tricorder in "Star Trek," a hand-held device or an app incorporated into a mobile device. After a quick and painless scan just by waving the device over the afflicted area, comes the prognosis—such as Barclay's protomorphosis syndrome, a pernicious DNA disease that makes one de-evolve in the most intriguing [science-fiction](#) way possible. While a device of this type remains science fiction, many of the medical tricorders' capabilities already exist in the form of bulky lab equipment and databases spread across the fields of omics.

Thorough investigation of disease, or disease susceptibility, requires a lot of different omics—genomics, epigenomics, transcriptomics, proteomics and metabolomics. The collection of multi-omics is costly and data-intensive, making it somewhat rare within research. These fields of knowledge are in databases full of detailed analyses of human cell functions and disease associations. Additionally, many multi-omics studies have been targeted at specific subset populations to interrogate disease mechanisms. A broader capture of multi-omics could confirm inferred knowledge and discover hidden biological pathways.

Xu and colleagues used a [machine-learning approach](#) to create genetic scores for 17,227 biomolecular traits from 48,813 healthy blood samples that can predict the levels of 13,668 RNA transcripts, 2,692 proteins and 867 metabolites. The genetic scores were then validated in seven diverse, independent cohorts.

The OmicsPred team anticipates that this new resource will be widely useful in investigating multi-omic traits and associations with biological traits. All the genetic scores from the study have been made public through the web portal omicspred.org, and the [scientific community](#) is invited to use it to predict multi-omic traits from genotypes in their own data sets.

Currently, the molecular traits that these scores predict only reflect the heritability and variability in the training data set, which was derived from healthy blood donors of predominantly white European ancestries. The creators plan to enhance and refine the range of genetic scores available in the OmicsPred resource and expand the ancestral diversity with updated training data sets.

More information: Yu Xu et al, An atlas of genetic scores to predict multi-omic traits, *Nature* (2023). [DOI: 10.1038/s41586-023-05844-9](https://doi.org/10.1038/s41586-023-05844-9)

Genome-based scores predict thousands of molecular traits in humans, *Nature* (2023). [DOI: 10.1038/d41586-023-00721-x](https://doi.org/10.1038/d41586-023-00721-x)

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