

Metabolomics-based predictors as markers of health and disease



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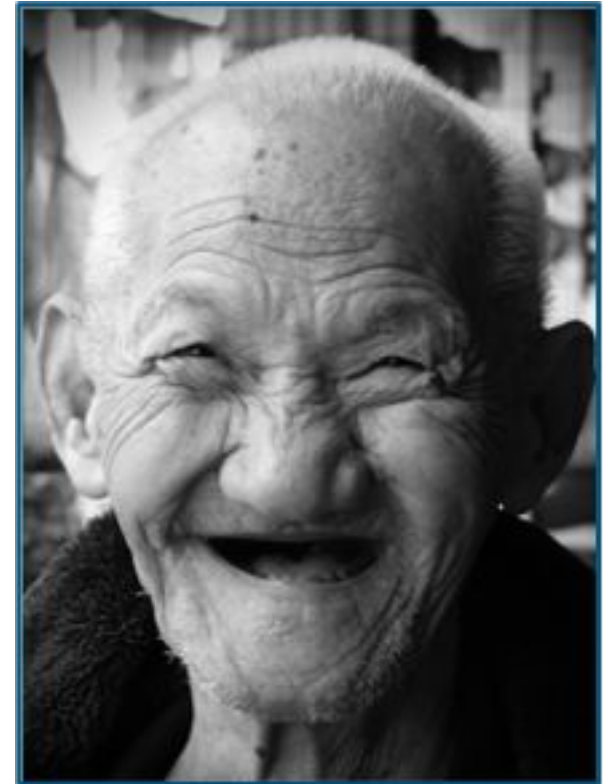
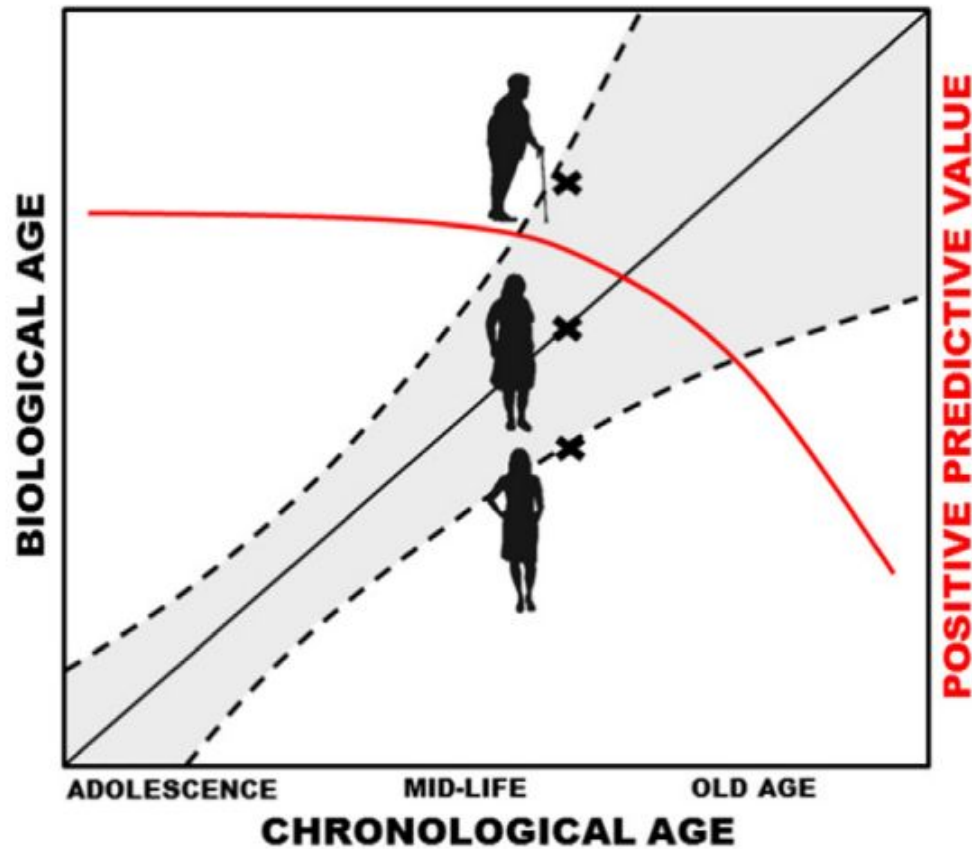
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Biological Age Biomarkers

To quantify the age-related decline



Contents lists available at ScienceDirect

EBioMedicine

journal homepage: www.ebiomedicine.com

ELSEVIER

EBioMedicine

Review

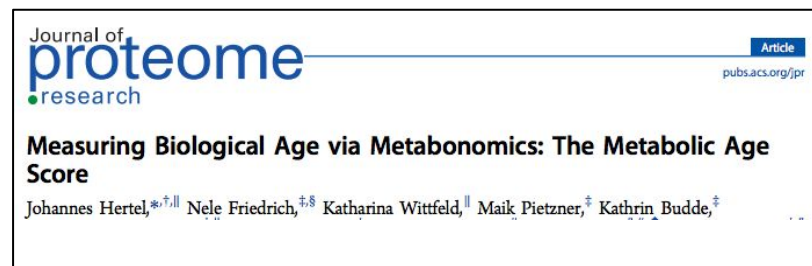
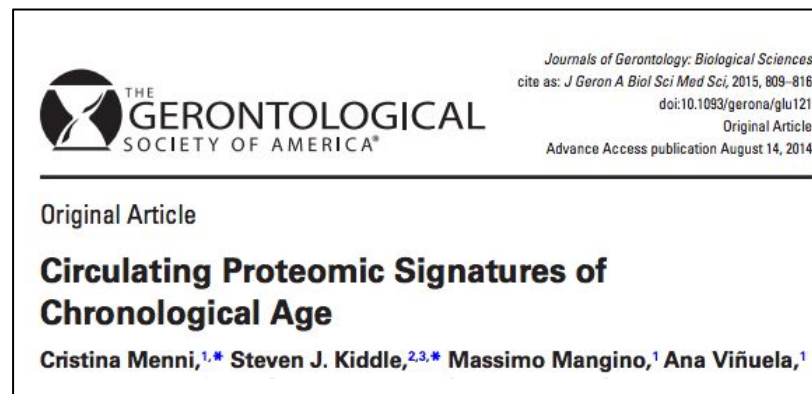
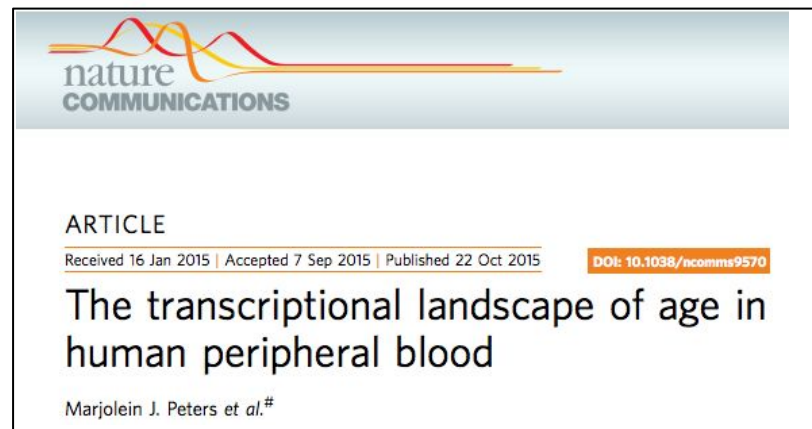
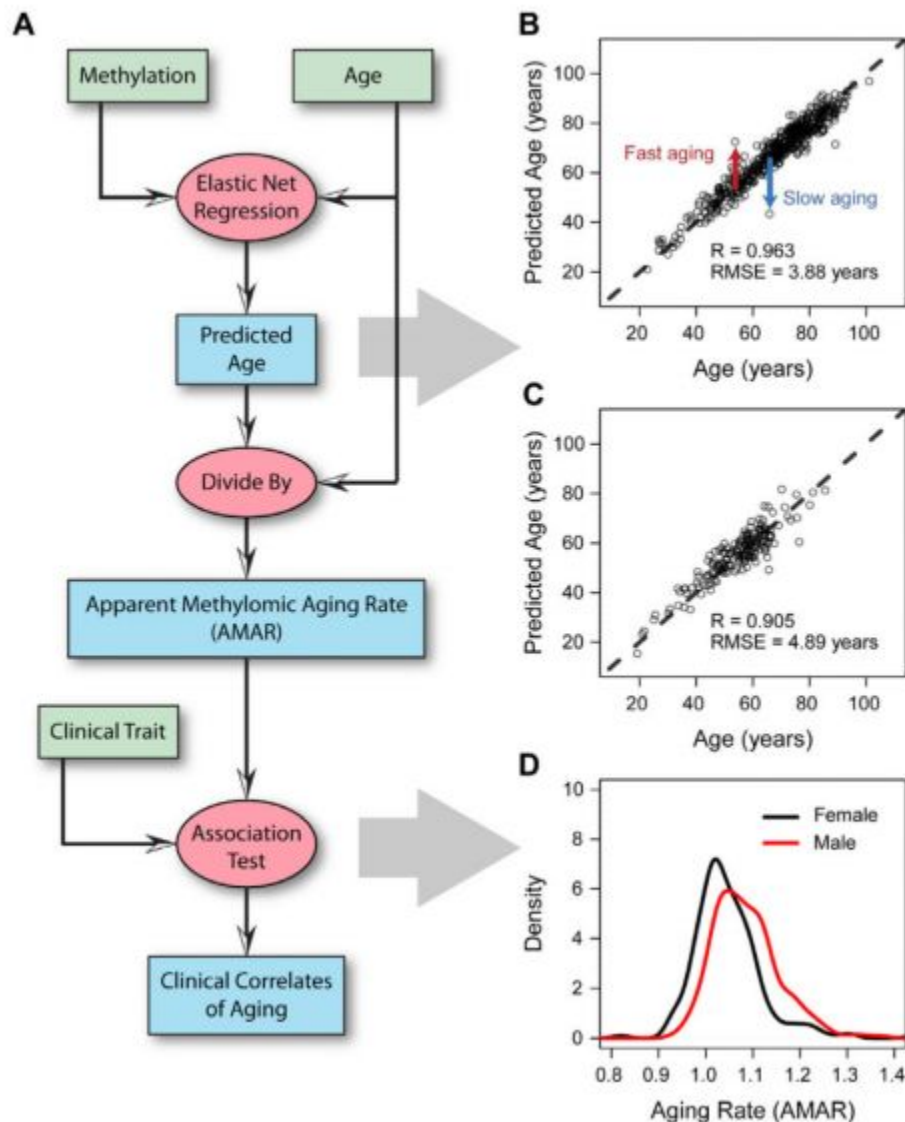
Biological Age Predictors

Juulia Jylhävä, Nancy L. Pedersen, Sara Hägg*

CrossMark

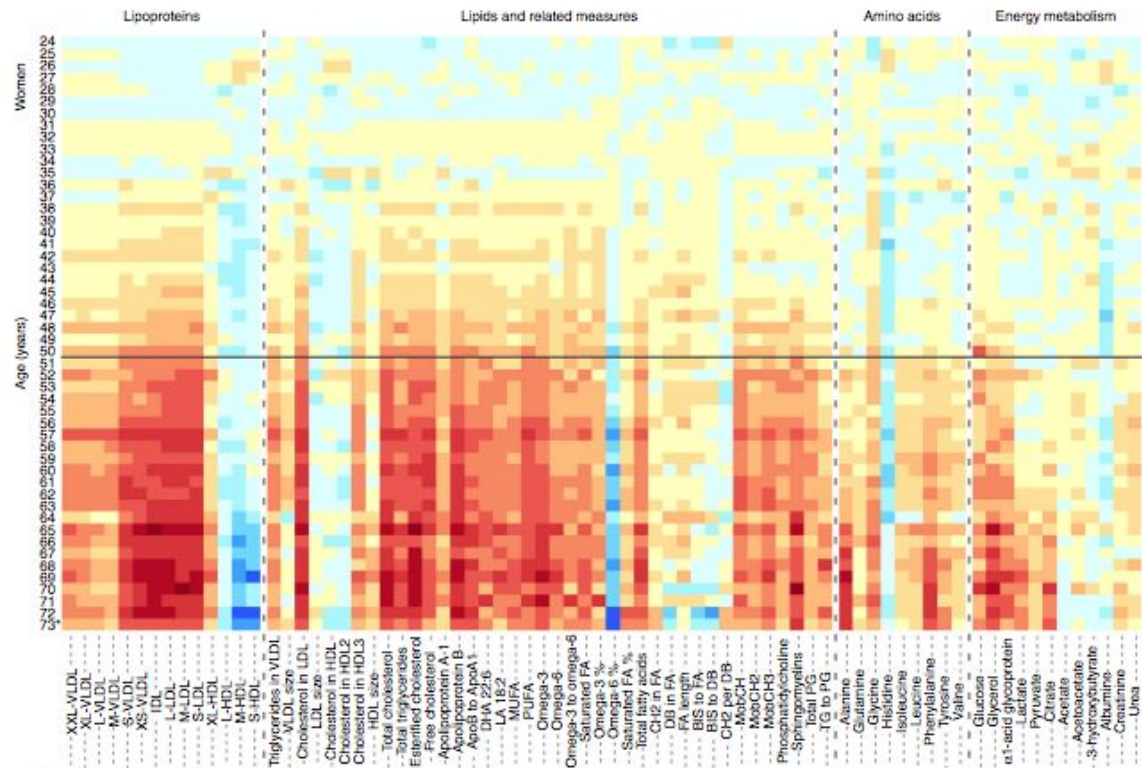
'Clocks'

The rationale for training (omics) age-predictors



Aim

Training and evaluating a Nightingale metabolomics age-predictor



Nightingale Health blood metabolomics

- ❑ Highly standardized
- ❑ Affordable
- ❑ 226 metabolites
- ❑ Soft cardiovascular focus
- ❑ Known age-related dynamics


nature
COMMUNICATIONS

ARTICLE

Received 22 Nov 2013 | Accepted 16 Jul 2014 | Published 21 Aug 2014

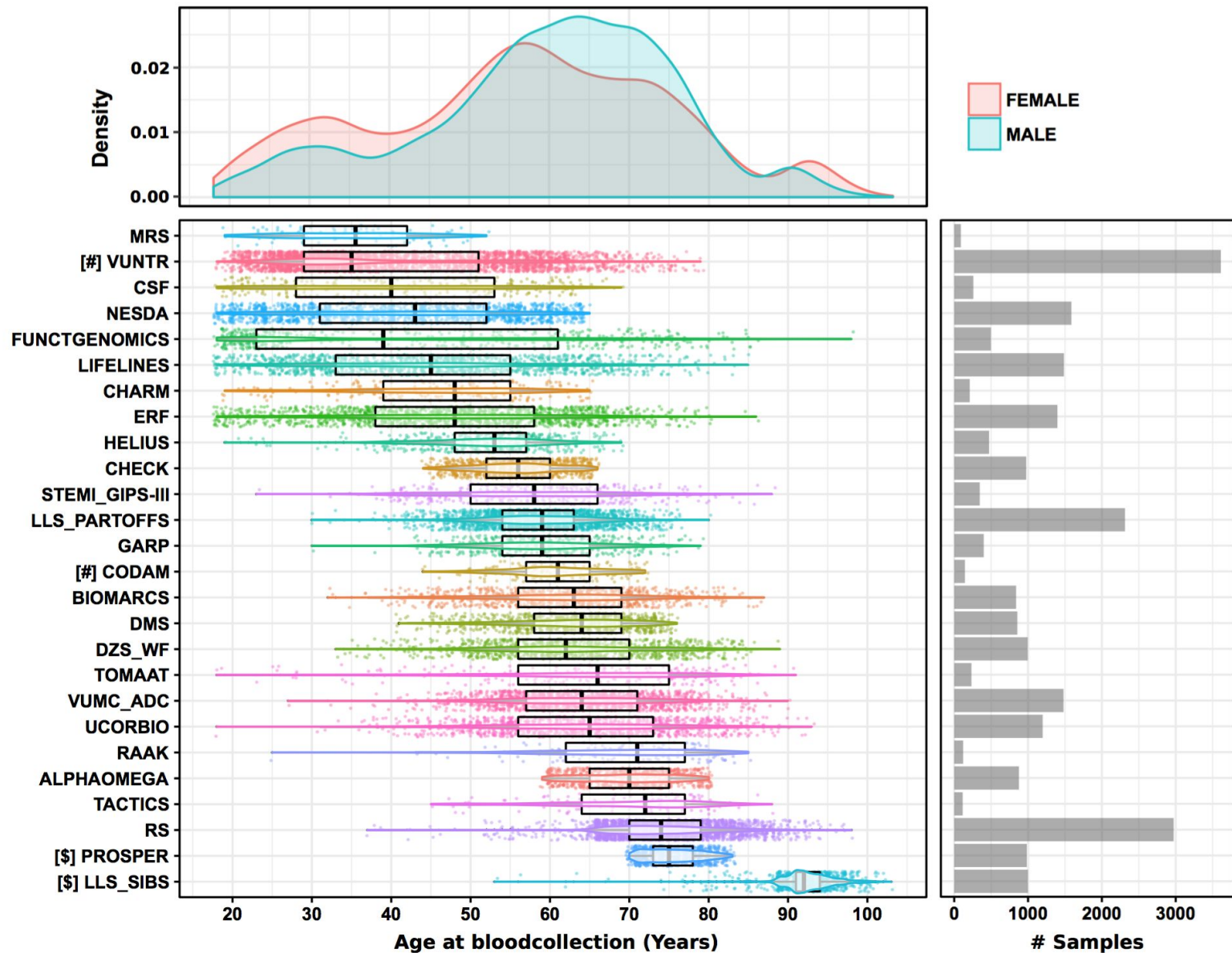
DOI: 10.1038/ncomms5708

A metabolic view on menopause and ageing

Kirsi Auro^{1,2,3,*}, Anni Joensuu^{1,2,*}, Krista Fischer⁴, Johannes Kettunen^{1,2,5}, Perttu Salo^{1,2}, Hannele Mattsson^{1,2},

Study Population after Quality Control

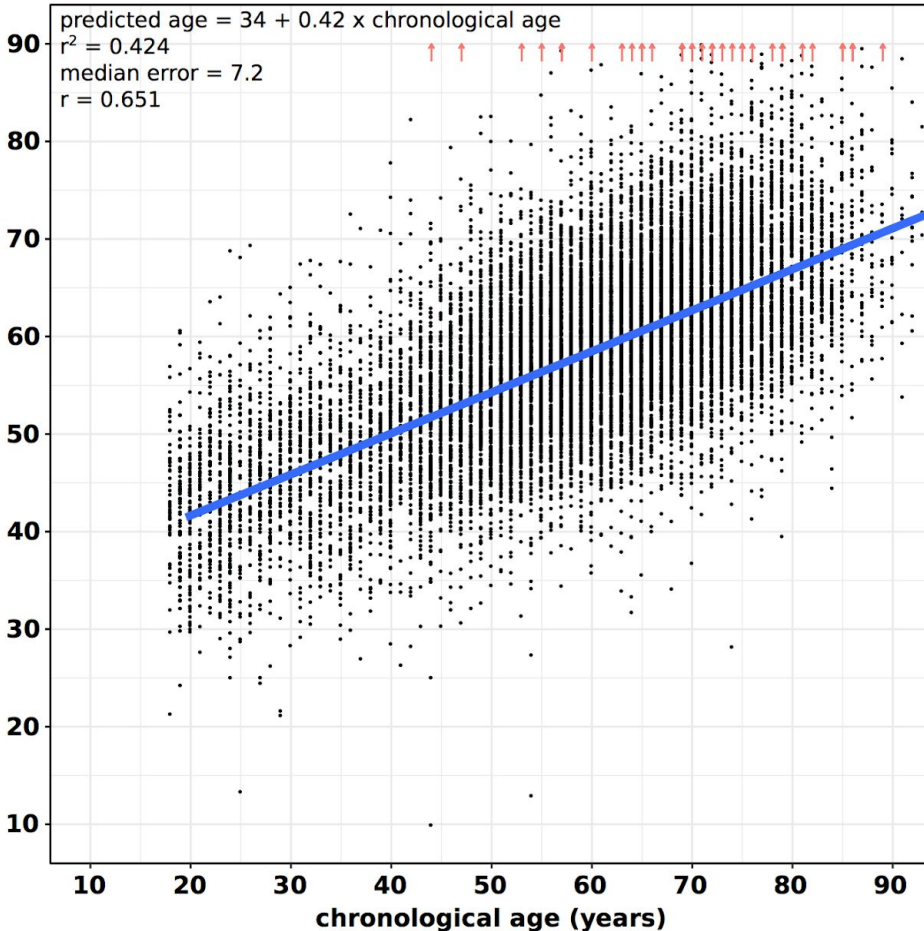
BBMRI: 25 453 samples from 26 biobanks



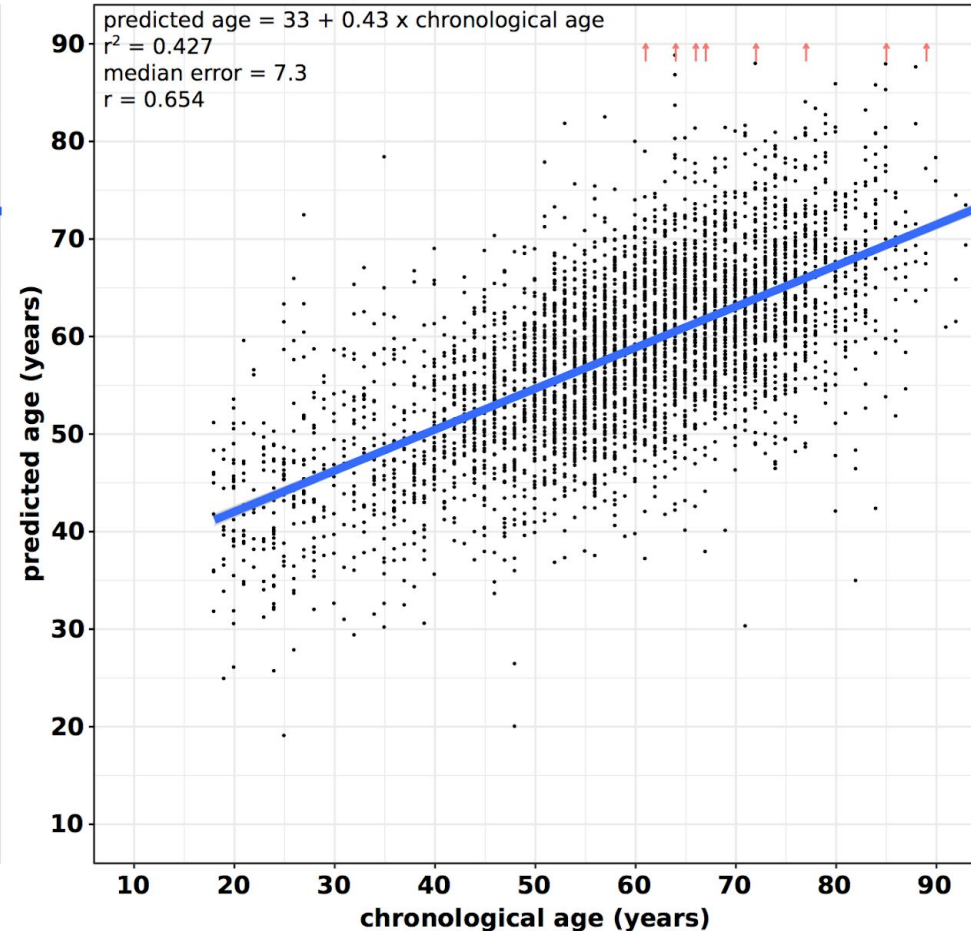
Training and evaluating the age-predictor

5 cross-fold validation

Training set (n = 15 208)



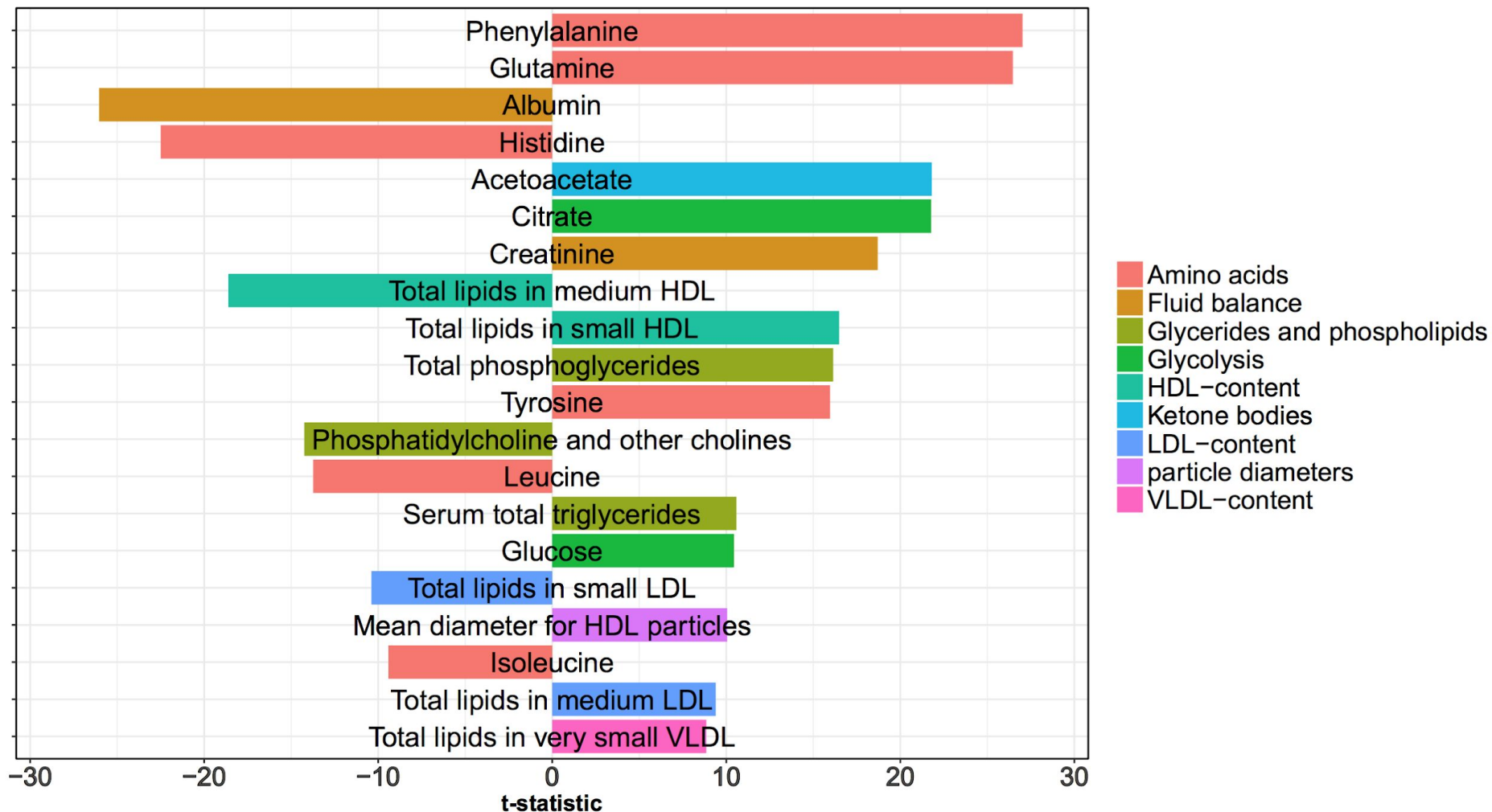
Test set (n = 3 802)



$$\text{predicted age} = \beta_0 + \beta_1 m_1 + \beta_2 m_2 + \dots + \beta_{56} m_{56} + \varepsilon$$

Training the age-predictor

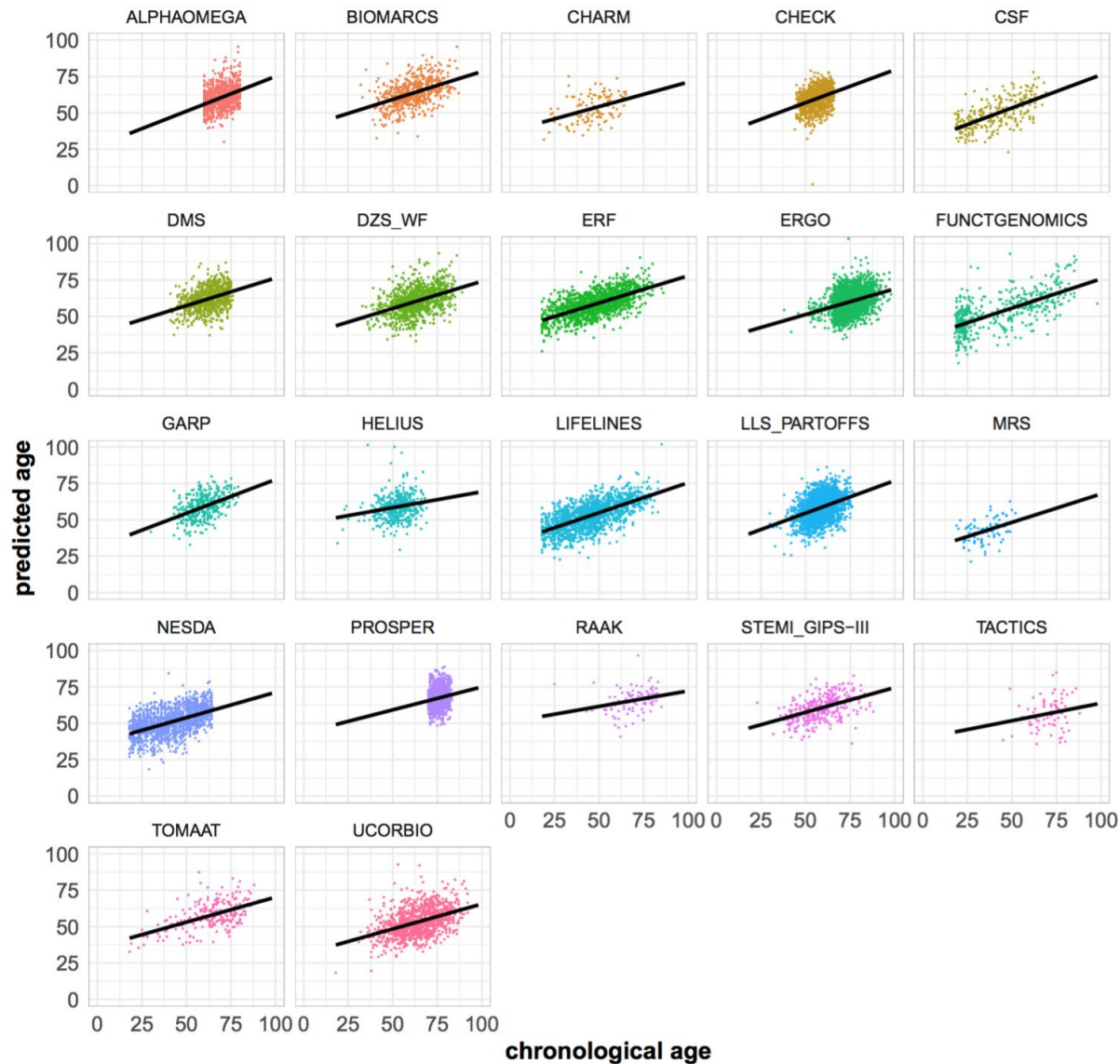
Contributing metabolites



Note: metabolites are scaled and centered before modelling!

Training and evaluating the age-predictor

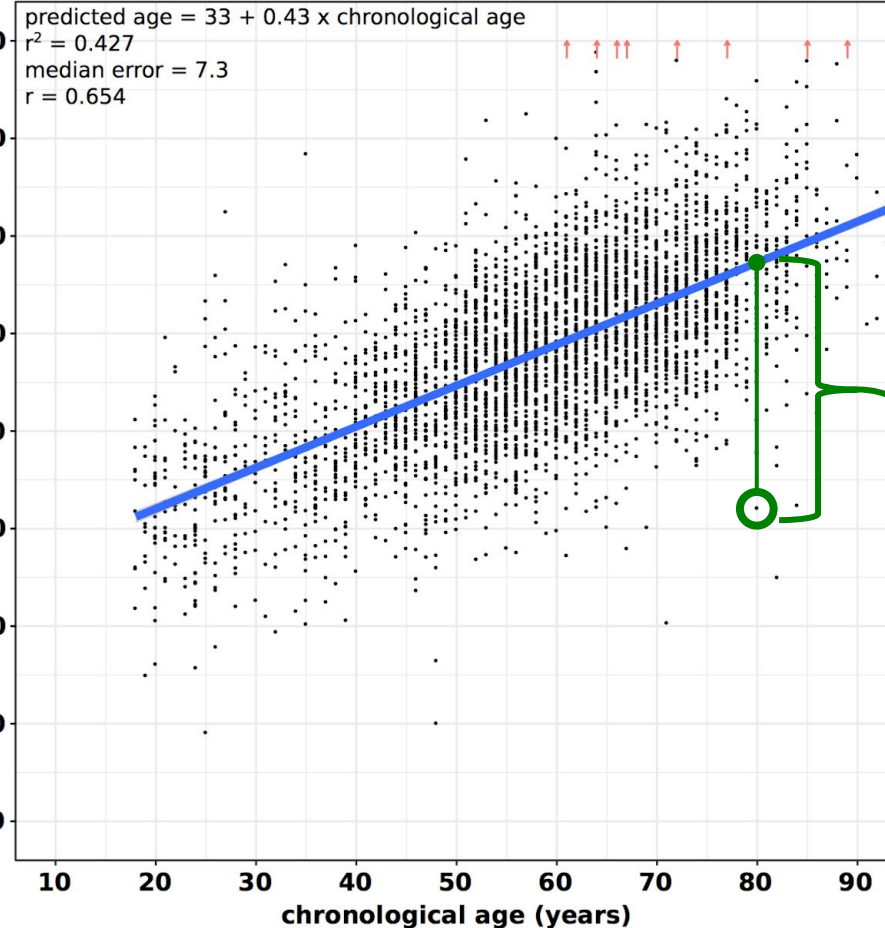
Leave-One-Biobank-Out Cross Validation



Construction of the metaboAge Score

The age-independent part of the age-predictor

Test set (n = 3 802)



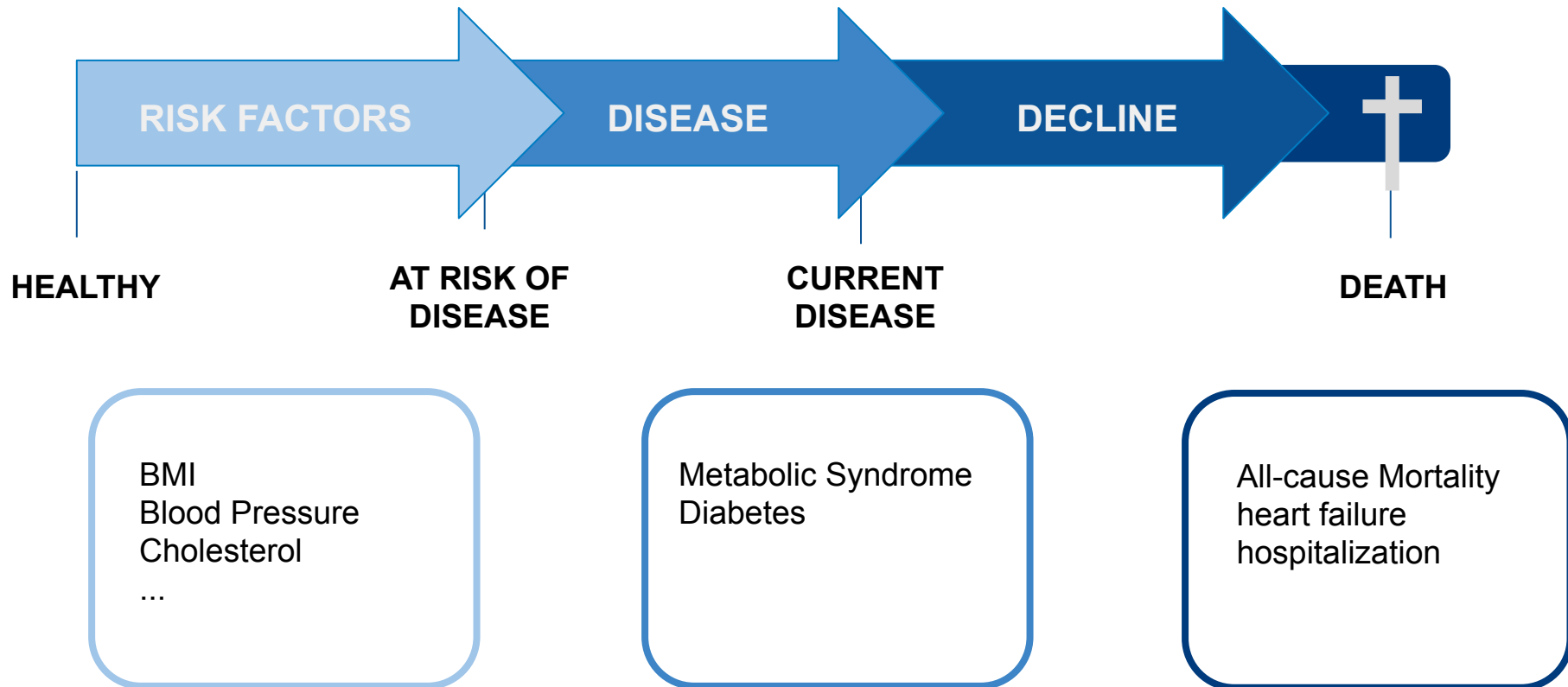
$\Delta \text{age} = \text{predicted age} - \text{chronological age}$

Regress out chronological age:

$\text{metaboAge} = \text{resid}(\text{lm}(\Delta \text{age} \sim \text{chron. age}))$

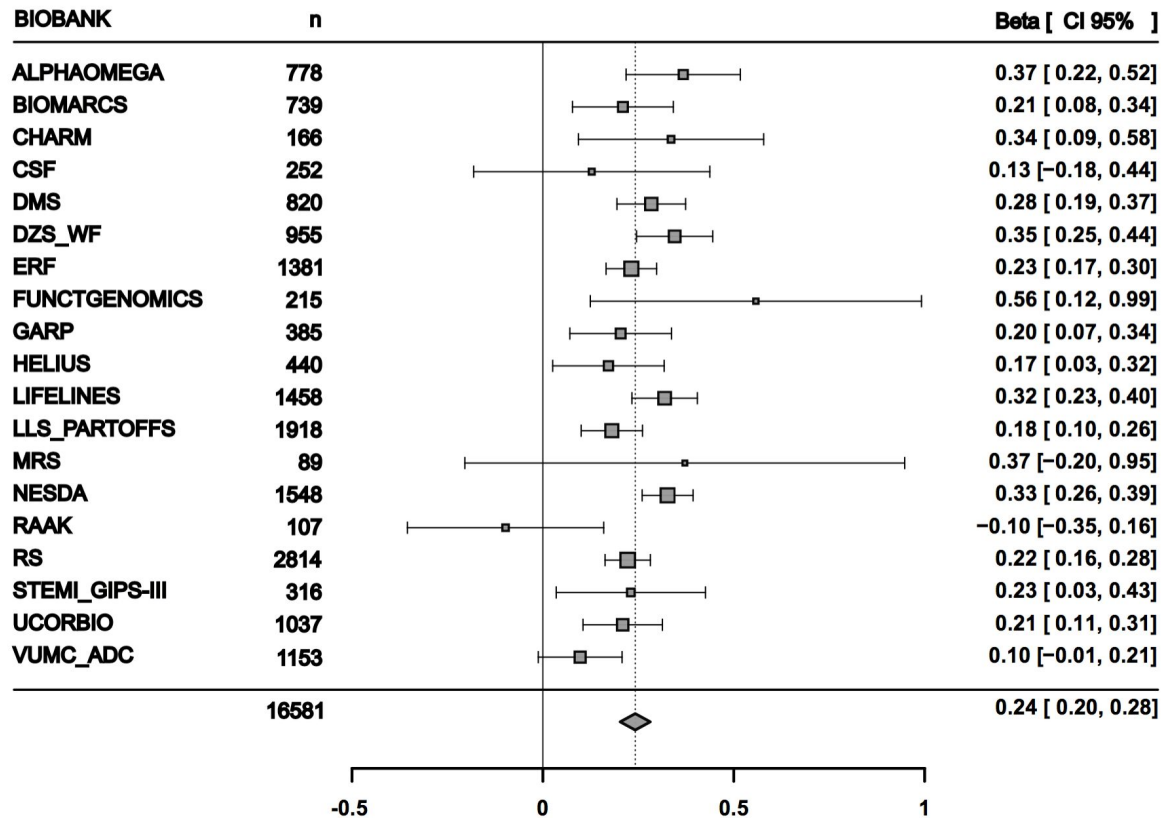
Evaluating the metaboAge Score

Tracking Age-Related Disease



Associations with Clinical Risk Factors

Body Mass Index



$$p_{meta} = 2.6 \times 10^{-33}$$

$$\Delta \text{age} = \beta_0 + \beta_1 \text{BMI} + \beta_2 \text{chronological age} + \beta_3 \text{gender} + \varepsilon$$



Significant associations

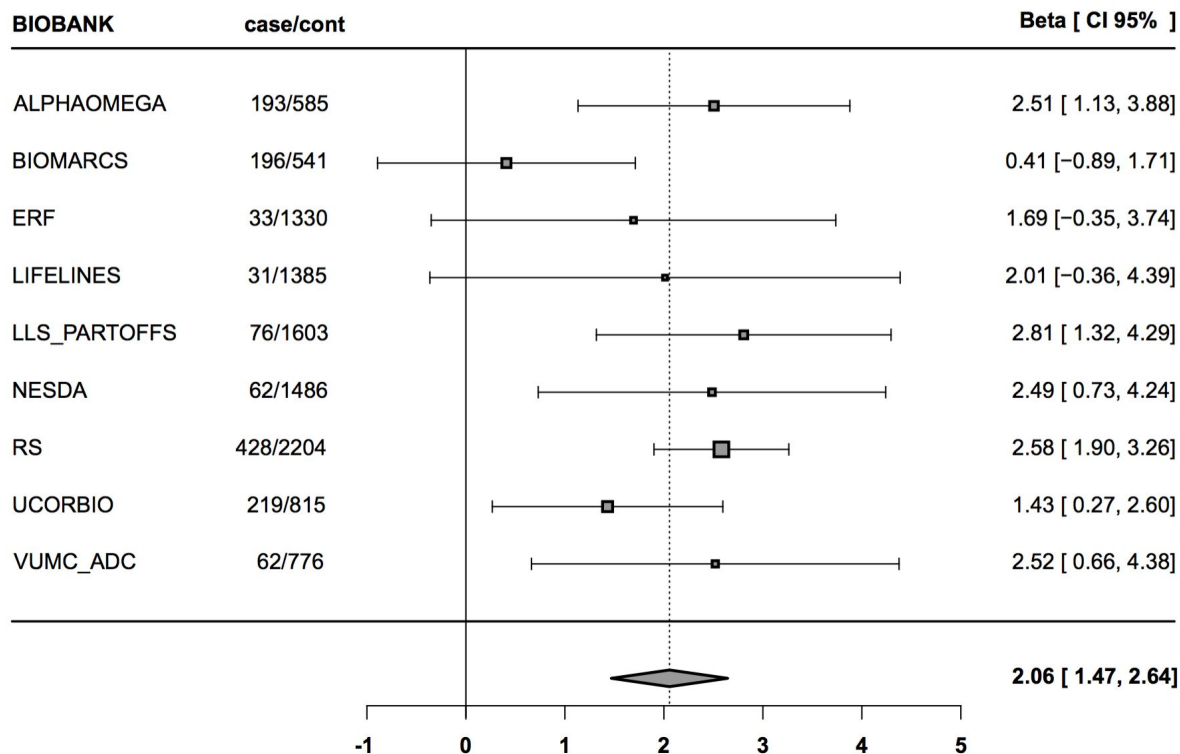
BMI
 hsCRP
 Total Cholesterol
 HDL Cholesterol
 Triglycerides
 Anti-hypertensive medication

No association

SBP
 DBP
 Lipid lowering medication
 Gender
 Smoking

Associations with Current Clinical Endpoints

Diabetes Mellitus Status



Significant associations

T2D

Metabolic Syndrome

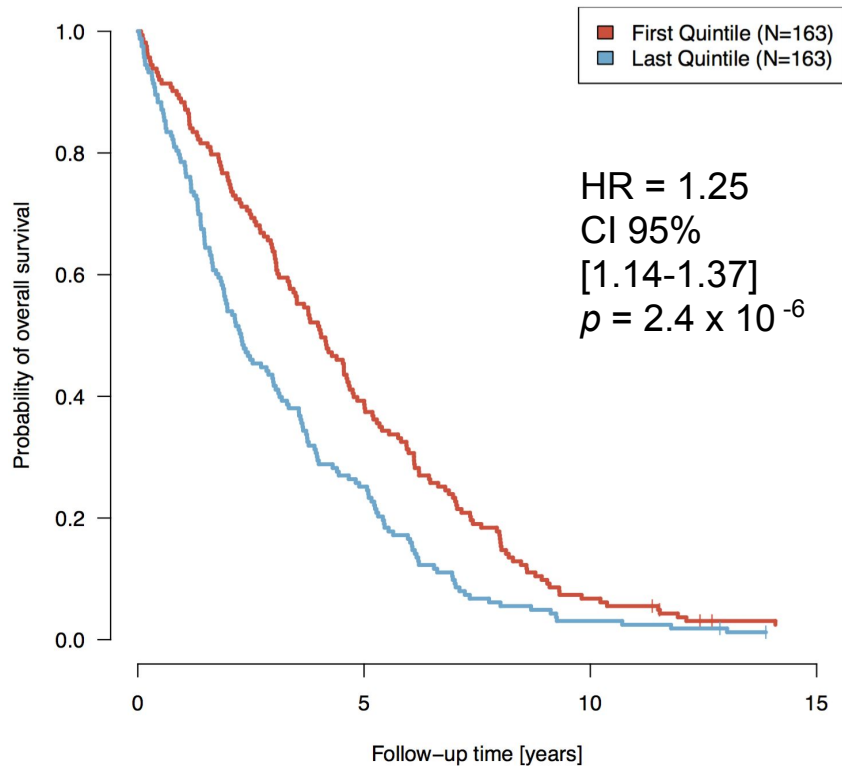
$$p_{meta} = 6.5 \times 10^{-12}$$

$$\Delta \text{age} = \beta_0 + \beta_1 \text{ diab status} + \beta_2 \text{ chronological age} + \beta_3 \text{ gender} + \beta_4 \text{ BMI} + \epsilon$$

Associations with Future Clinical Endpoints

Mortality / Cardiovascular events

LEIDEN LONGEVITY STUDY - MORTALITY



Cox proportional hazards model:

$$FUT = \beta_0 + \beta_1 \Delta \text{age} + \beta_2 \text{chronological age} + \beta_3 \text{gender} + \varepsilon$$

PROSPER - INCIDENT CARDIOVASCULAR EVENTS

Phenotype	HR	95% CI	p-value
Coronary events	1.25	1.11 – 1.40	2.64×10^{-04}
Cardiovascular events	1.20	1.08 – 1.33	4.86×10^{-04}
Vascular mortality	1.57	1.31 – 1.88	8.56×10^{-07}
All-cause mortality	1.42	1.25 – 1.61	9.14×10^{-08}
Heart failure hospitalisation	1.68	1.37 – 2.06	5.42×10^{-07}



Effect independent of BMI, smoking, diabetes and hypertension and medication

Conclusions

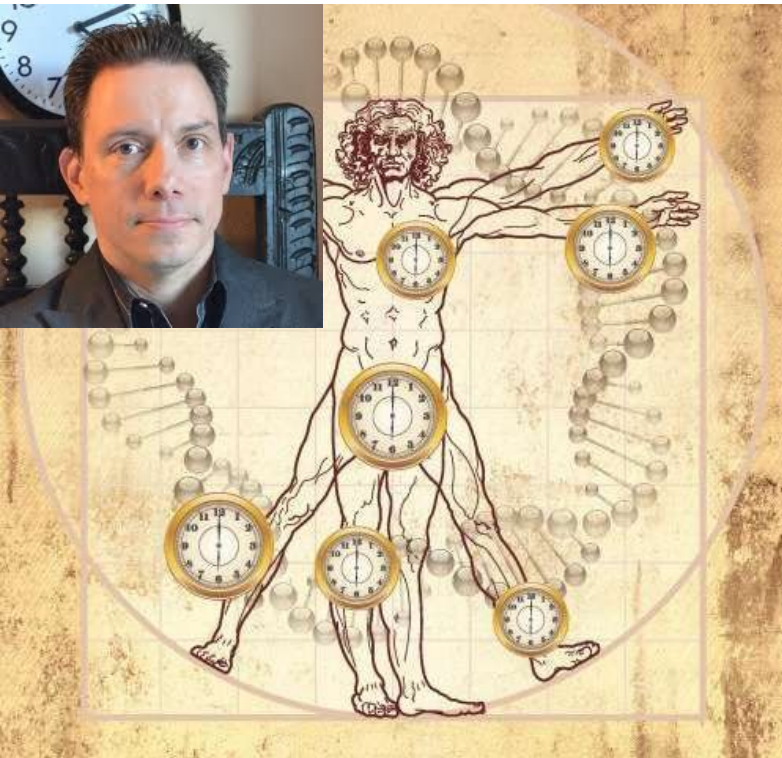
Summary of main findings

- *Nightingale Health* blood metabolomics predicts age reasonably ($r = 0.65$)
 - Comparable with transcriptomics & metabolomics clocks
 - Worse than epigenetics clocks ($r > 0.9$). **Does this matter?**
- *MetaboAge* shows promise as an biomarker
 - *MetaboAge* associates significantly with clinical risk factors
 - *MetaboAge* associates significantly with current clinical end points
 - *MetaboAge* associates significantly with future clinical end points
- *MetaboAge* is:
 - Highly standardized
 - Affordable
 - Easy to implement

Discussion

Other clocks

Small correlation with Horvath's methylation clock



What outcomes do the clocks predict?
What do they predict independently?

Small correlation with Deelen's mortality predictor



ARTICLE

<https://doi.org/10.1038/s41467-019-11311-9>

OPEN

A metabolic profile of all-cause mortality risk
identified in an observational study of 44,168
individuals

Joris Deelen  et al.[#]

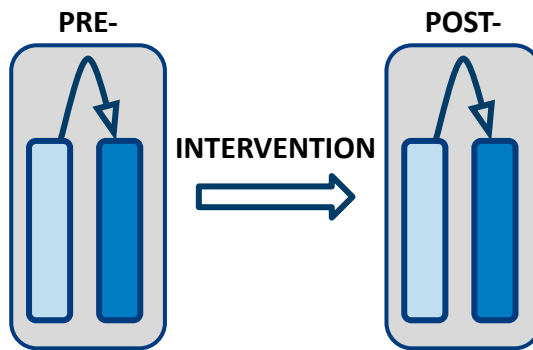
Discussion

Metabolomics as a read-out of intervention success



Fatih Boogards

MetaboAge is not affected by the intervention
Yet the intervention can be captured by metabolomics predictors



IN SHORT: NUMBERS

164 middle-aged
undergoing a **13-week**
lifestyle intervention

12.5% less intake
12.5% more exercise

Timepoint [t]

1

2

3

4

Muscle



→ RNAseq; Muscle Parameters [*in progress*]

Blood



→ RNAseq; **Metabolomics**; Classical Parameters

Adipose



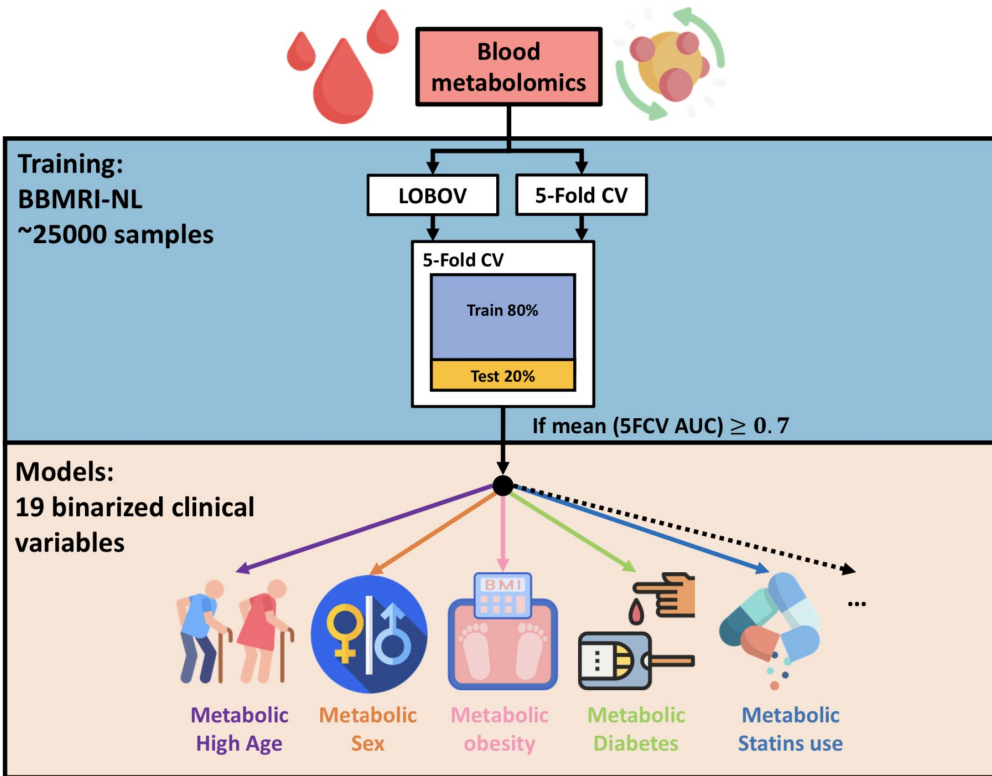
→ DEXA; Genetics; FFQ; Accelerometer; ...

Etc ...

Outlook & related work

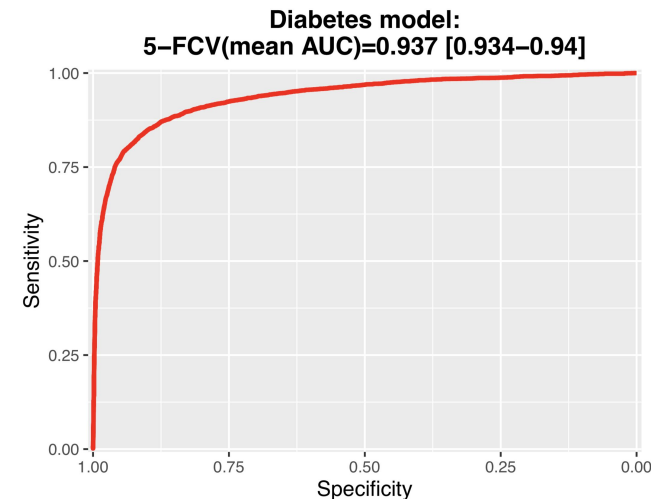
Clock-like approaches: other surrogate markers

Other risk factors might be readily predicted ...
... And transformed in discrepancy-based predictors



Daniele Bizzarri

Application 1: Imputation



Application 2: Risk prediction

Δ Diabetes, Δ Obesity, ...

Outlook & related work

Clock-like approaches: what platform captures relevant signal?



Cathelijn Kuijt

Dennis Mook

Platform	Samples	Measurements
<i>Nightingale</i>	349	224
<i>Biocrates</i>	356	163
<i>Lipidyzer</i>	307	1107
<i>Metabolon</i>	304	762

ELANET (fixed) <i>R-squared</i>	AGE		BMI		SEX		DIAB	
	RSQ	df	RSQ	df	RSQ	df	RSQ	df
<i>Nightingale</i>	35,72%	96	36,01%	42	68,91%	58	47,88%	45
<i>Biocrates</i>	47,25%	121	42,17%	96	55,96%	94	55,49%	50
<i>Lipidyzer</i>	51,88%	154	17,51%	104	38,03%	231	35,75%	129
<i>Metabolon</i>	59,50%	162	48,66%	188	77,00%	195	60,23%	135

Some platforms are more predictive for cardio-metabolic variables

metaboAge

Welcome to the webtool for calculating metabolic age (*'metaboAge'*) from raw Nightingale Health 1H-NMR metabolomics data.

Please refer to our manuscript when using metaboAge in your work:

E.B. van den Akker et al. "MetaboAge: a novel biomarker for biological age based on the BBMRI-NL 1H-NMR metabolomics repository" ([submitted](#))



Jurriaan Barkey Wolf

Instructions

For the webtool to work, the input data needs to be provided in a specific format. This format should be similar to the raw data files you received from Nightingale Health.

Please ensure that your dataset looks like this [example dataset](#) before submitting.

This means that:



Daniele Bizzarri

R-shiny application is coming to compute metaboAge and other scores!

Acknowledgements



Section of Molecular Epidemiology

Eline Slagboom

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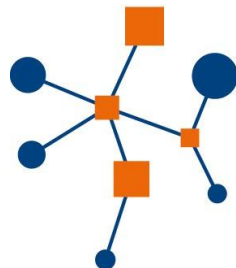
Thies Gehrman

Daniele Bizzarri

Jurriaan Barkey Wolf

Joris Deelen

Fatih Bogaards



BBMRI.nl

Biobanking and
BioMolecular resources
Research Infrastructure
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Dennis Mook

Stella Trompet

Leiden Computational Biology Center

Marcel Reinders

Cathelijn Kuijt

BBMRI-NL Consortium Partners



Netherlands Consortium for
Healthy Ageing

