

multi-ancestry Polygenic Risk scOres based on enSemble of PEnalized Regression models (PROSPER)

Jingning Zhang*, Jianan Zhan, Jin Jin, Cheng Ma, Ruzhang Zhao, Jared O' Connell,

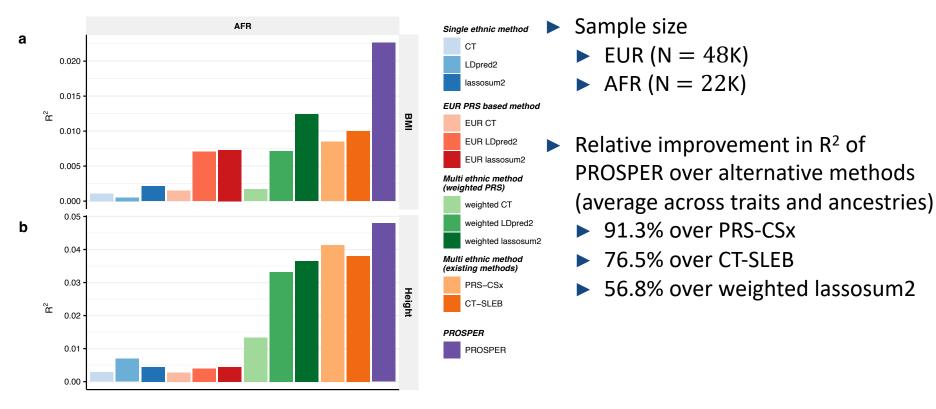
Yunxuan Jiang, 23andMe Research Team, Bertram L. Koelsch, Haoyu Zhang, Nilanjan Chatterjee*

*Correspondence to: Jingning Zhang (jzhan218@jhu.edu) and Nilanjan Chatterjee (nilanjan@jhu.edu)

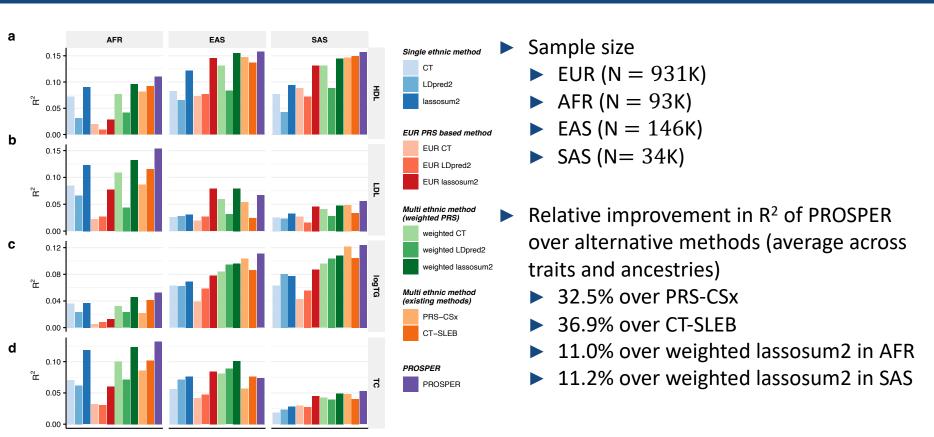
The proposed method -- PROSPER

Details of the proposed model will be released right after the manuscript being submitted to journal

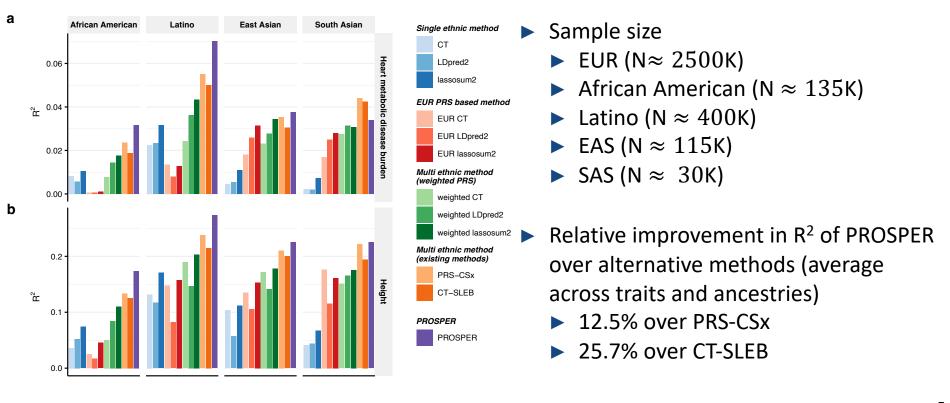
Results on data from All of US (AoU)



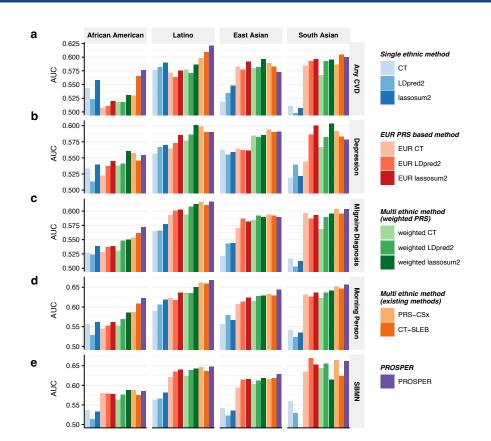
Results on data from Global Lipids Genetics Consortium (GLGC)



Results on data from 23andMe Inc. (23andMe) continuous traits



Results on data from 23andMe Inc. (23andMe) binary traits



- Sample size
 - ► EUR (N≈ 2370K)
 - ► African American (N ≈ 110 K)
 - ► Latino (N ≈ 400 K)
 - ► EAS (N ≈ 86K)
 - ► SAS (N \approx 24K)
- Relative improvement in AUC of PROSPER over alternative methods (average across traits and ancestries)
 - ► 1.1% over PRS-CSx
 - ▶ 1.3% over CT-SLEB

Runtime and memory usage

► Chromosome 22

two ancestries: AFR, and EUR

five ancestries: AFR, AMR, EAS, EUR, and SAS

► Runtime and memory usage for model training part

Method	Computational time (minutes)	Memory (Gb)
PROSPER (two ancestries)	3.0	2.24
PROSPER (five ancestries)	6.8	2.35
PRS-CSx (two ancestries)	111.1	0.78
PRS-CSx (five ancestries)	595.8	0.84

Conclusion

- PROSPER has substantial improvement over alternative methods and can deal with traits with a variety of genetic architectures
- ▶ PROSPER is not only more powerful for complex traits with high polygenicity, but also robust to biomarker traits with large-effect loci. In addition, PROSPER developed based on penalized regression is an order of magnitude faster compared to alternative Bayesian methods.