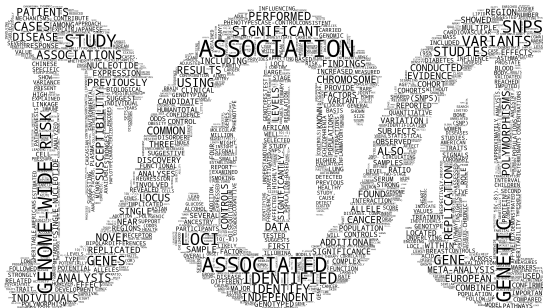


Monitoring GWAS Diversity in Pseudo-Real Time

M.C. Mills, **C. Rahal**, E. Akimova, D. Leasure et al.

Leverhulme Centre for Demographic Science and Nuffield College

Link23 Launch, 29th March, 2023



What is a **Genome Wide Association Study**?

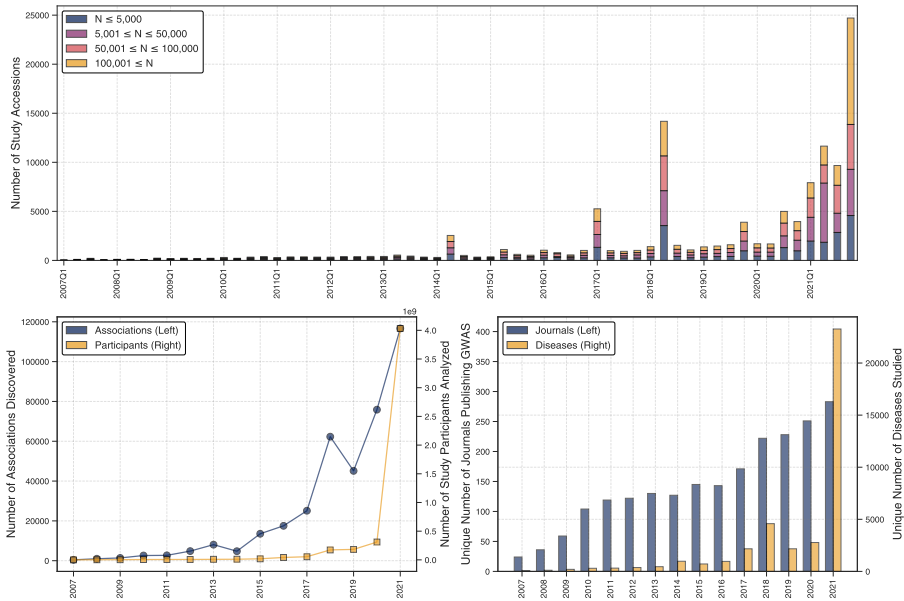
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- A typical GWAS will have both a *discovery* and *replication* sample.
 - Discovery sample summary statistics generate PRS for clinical purposes.
- Specific examples include:
 - **Colorectal cancer**: UK Biobank, 26.3%- 30.1% variance explained.
 - **Schizophrenia**: up to 22% of variance explained.



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- Largest Accession: **4.1m** people (human height, PMID: 36224396).

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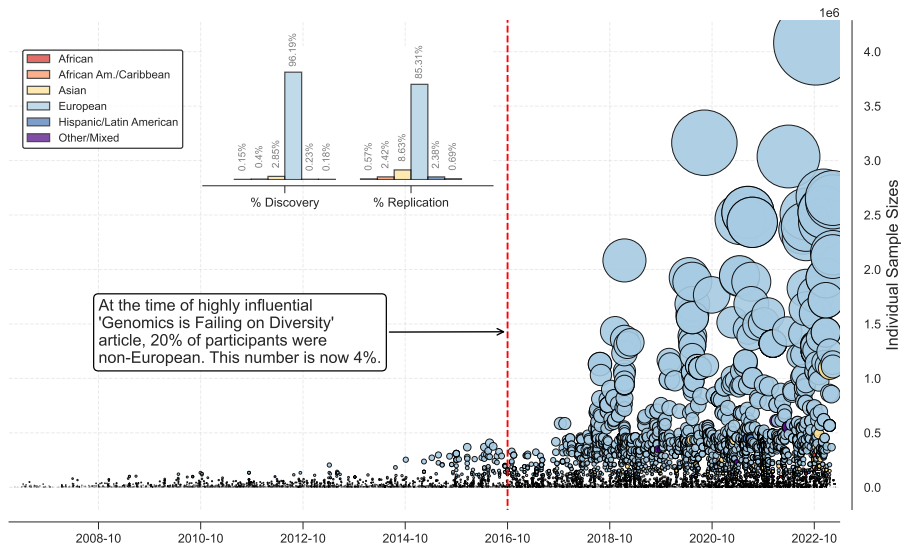
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- Public health debate about 'personalized medicine': personalised for who?
- Analysis of diverse populations **also** improves discovery for complex traits!

Majority of studies are on people of European Ancestry



This is all made possible by **amazing** data originators.



Introducing the GWAS Diversity Monitor

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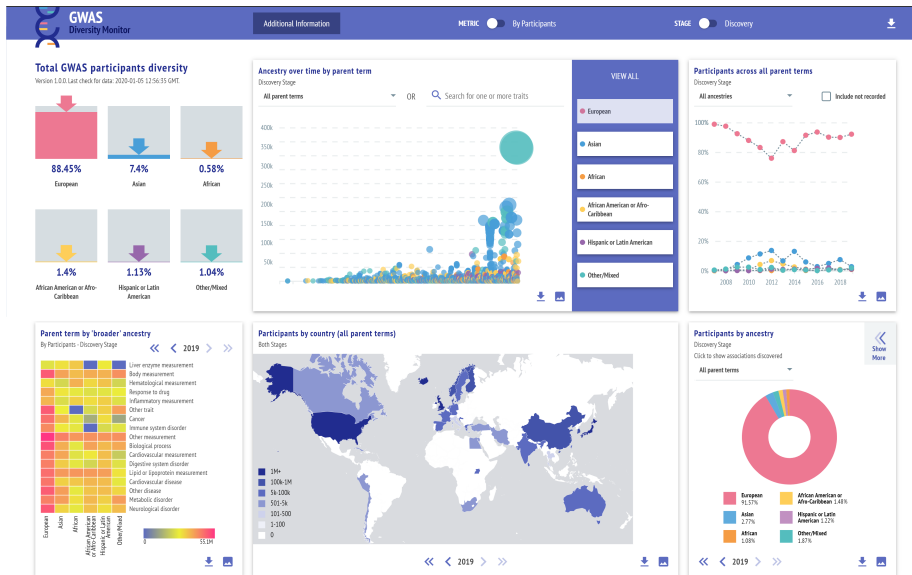
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Thanks to all the Funders and (much) Broader Team!



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