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10th November 2023 DRI Workshop: Molecular Epidemiology and Big Data for Dementia Research

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Motivation:

Genetic evidence for drug-target discovery and validation Single-cell sequencing to identify cell-types of action

Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR) Single-cell eQTL mapping Single-cell MR

Results: Single-cell expression affecting human brain disease and behaviour

Conclusion and Outlook

- Motivation:

Genetic evidence for drug-target discovery and validation

Genetic evidence for drug-target discovery and validation

- Most drugs fail during clinical development
- Most common reason for this is lack of efficacy in Phase II/III trials due to inadequate target validation in early-stage drug discovery
- Of 216 new drugs entering German market between 2011-2017, 75% showed no quantifiable benefit in efficacy over existing drugs for the same indication
- For neurology/psychiatry indications 94% provided no added efficacy benefit over existing therapies

[Wieseler et al., 2019]

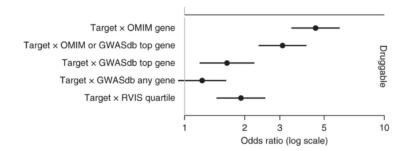
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- Motivation:

Genetic evidence for drug-target discovery and validation

Genetic evidence for drug-target discovery and validation

 Genetic evidence is increasingly used for drug-target discovery, prioritization and validation



[Nelson et al., 2015]

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- Motivation:

Single-cell sequencing to identify cell-types of action

Single-cell sequencing to identify cell-types of action

- Single-cell RNA sequencing (scRNA-seq) allows to measure the transcriptome at single-cell resolution and has given insights into gene-regulation of different cell-types
- In contrast standard bulk-tissue measurements average expression of a transcript over all cell-types
- Linking genotype with scRNA-seq data allows novel insights into genetic regulation of cell-type specific effects

[Cuomo et al., 2023]

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- Motivation:

Single-cell sequencing to identify cell-types of action

How to link cell-type specific gene-expression with disease and behavioral outcomes?

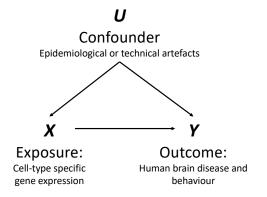


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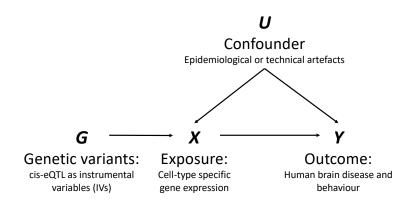


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- Motivation:

Single-cell sequencing to identify cell-types of action

Two-sample summary-level Mendelian randomization (MR)

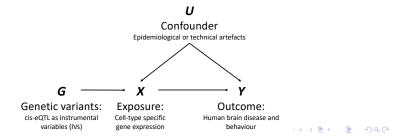


- Motivation:

Single-cell sequencing to identify cell-types of action

The small print: Instrumental variable (IV) assumptions

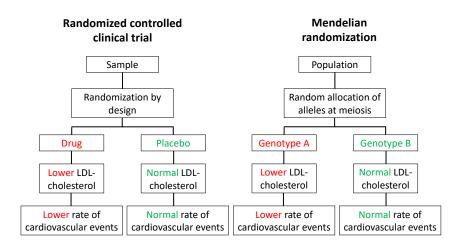
- If all genetic variants used as IVs are valid, MR can estimate the true causal effect unbiased from any confounders U.
- ► A genetic variant is a **valid IV** if the following criteria hold:
 - IV1 Relevance: The variant is associated with the exposures
 - IV2 **Exchangeability**: The variant is independent of the confounders U of the exposure-outcome associations
 - IV3 Exclusion restriction: The variant is independent of the outcome Y conditional on the exposure X and confounder U



- Motivation:

Single-cell sequencing to identify cell-types of action

Analogy: Randomized controlled clinical trial and MR



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└─ Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

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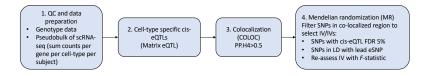
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Conclusion and Outlook

-Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

Analysis plan



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└─ Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

Single-cell eQTL mapping

Pre-processing

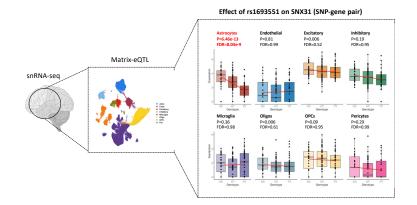


- Post QC n=128 with 4,500 cells/subject
- 587k nuclei across the sample set
- Integrated, clustered and cell-types annotated using canonical markers (8 major cell-types)

– Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

Single-cell eQTL mapping

Single cell-type cis-expression quantitative trait loci (eQTL) mapping

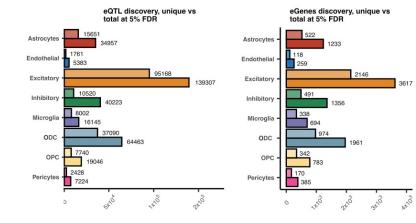


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- Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

Single-cell eQTL mapping

Single cell-type cis-eQTL mapping

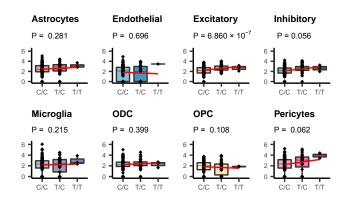


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Dash Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

Single-cell eQTL mapping

Genetic regulation of cell-type specific effects rs1716183 / OGFOD2

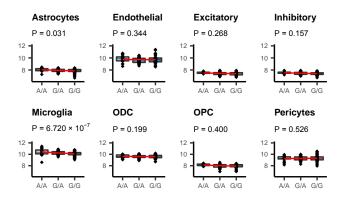


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– Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

Single-cell eQTL mapping

Genetic regulation of cell-type specific effects rs10792832 / PICALM

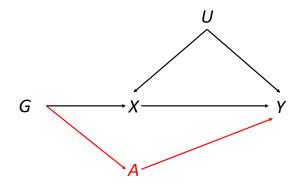


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Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

Single-cell MR

MR and genetic confounding



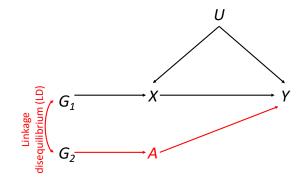
There is no pleiotropic pathway *A* that directly connects *G* with *Y*.

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– Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR)

└─Single-cell MR

Drug-target/cis-MR and genetic confounding



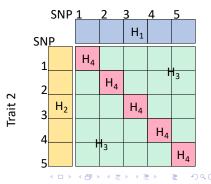
 Colocalisation is a necessary filtering step to ensure cis-MR results are not confounded by LD [Zuber et al., 2022] Single-cell Mendelian randomisation on human brain disease and behaviour Methods: Single-cell expression quantitative trait loci (eQTL) mapping and Mendelian randomization (MR) Single-cell MR

Colocalisation

Bayesian approach to test if two traits share the same genetic architecture in a region of interest [Giambartolomei et al., 2014]

Trait 1

- H₁: Association with trait 1
- H₂: Association with trait 2
- H₃: Association with trait 1 and 2, but different SNP
- H₄: Association with trait 1 and 2, one shared SNP



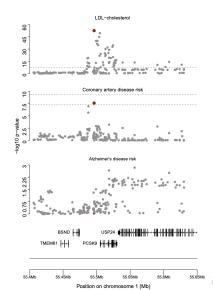
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└─ Single-cell MR

Colocalization to support MR

PCSK9 inhibitors

- PCSK9-gene region: PCSK9 is an enzyme that binds to and degrades the receptor for low-density lipoprotein particles (LDL)
- Exposure: LDL-cholesterol
- Primary outcome: Coronary artery disease
- Secondary outcome: Alzheimer's disease
 [Zuber et al., 2022]



Motivation:

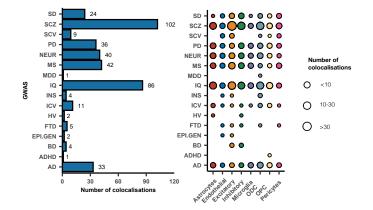
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Colocalization for the selection of valid IVs



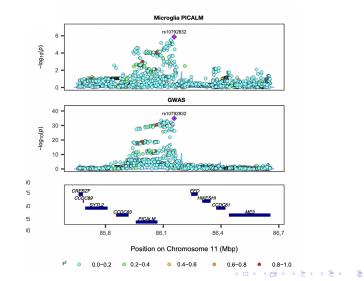
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Colocalization example: Alzheimer's Disease (AD)

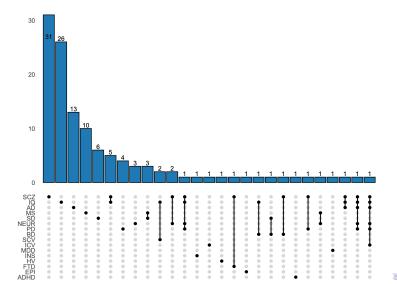
Alzheimer's Disease Astrocytes Endothelial Excitatory Inhibitory Microglia ODC OPC Pericytes										
		es Endothella	al Excitator	y inhibitory	/ Microglia	ODC	OPC	Pericytes		
SNX3	1 - 0.99	0.06	0.39	0.19	0.05	0.45	0.13	0.06		
RIN		0.12	0.06		0.99			0.05		
PICAL			0.07	0.09	0.99	0.08	0.06	0.03		
LRRC37/		0.13	0.99	0.99	0.31	0.3	0.36	0.23		
JAZF			0.08	0.06	0.99	0.07	0.12	0.07		
NSI		0.1	0.94	0.76	0.18	0.95	0.83	0.2		
EGF		0.06	0.09	0.06	0.08	0.05	0.08	0.05		
CR		80.0	0.06	0.16	0.07	0.92	0.73	0.11		
RASGEF10		0.07	0.08	0.09	0.87	0.08	0.09	0.1		
LRRC37A			0.33	0.82	0.37	0.28	0.34	0.35		
d TRIM7:	2 - 0.07	0.06	0.06				0.06	0.78		
eue CR			0.04	0.06	0.06		0.76	0.11		
110000		80.0	0.67	0.22	0.08	0.07	0.06	0.07		
RABEP				0.15	0.35	0.66	0.18	0.07		
ATP5F1		0.06		0.06	0.17	0.06	0.65	0.06	PP.F	
CFAP12		0.06		0.06	0.06	0.63	0.06	0.07		1.00
DPY19L		0.09			0.06	0.06	0.06	0.63		0.75
YPEL		0.07	0.15		0.18	0.26		0.06		0.50
CLI		0.59	0.09	0.06	0.06	0.05		0.2	-	0.25
ADAM1		0.06	0.06			0.59	0.08	0.08	-	0.00
FCER10		0.07				0.06		0.08		
TRPV					0.07	0.06		0.58		
CD2AI		0.06	0.09	0.06	0.53	0.39	0.08	0.06		
PPP40		0.06	0.06		0.18	0.52	0.06	0.07		
CNOT		0.06				0.13		0.07		
POP		0.09	0.06	0.18		0.52	0.06	0.09		
OPN		0.51	0.03			0.06		0.05		
MINDY	2 - 0.06	0.08	0.1	0.07	0.03	0.51	0.06	0.1		

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Colocalization example: AD and PICALM in Microglia

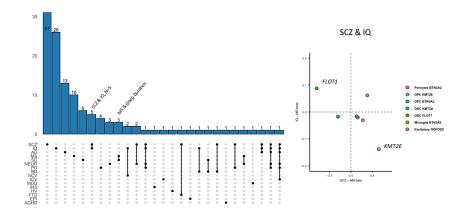


MR results: Overview



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MR results for more than one phenotype: SCZ and IQ



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MR-Results: Key highlights

			MR	MR	eQTL
Outcome	Gene	Cell-type	p-value	beta	FDR
AD	PICALM	Microglia	3.03E-36	Negative	6.26E-08
AD	EGFR	Astrocytes	1.70E-07	Positive	0.001
Parkinson	GPNMB	Astrocytes	3.01E-06	Positive	0.03

- PICALM modulates AD risk via clearance of Aβ and tau (Van Acker at al., Molecular Neurodegen 2019)
- Reduction in *PICALM* expression increases tau deposition. (Ando et al., Acta Neuropath 2020)
- Suggests *PICALM* activation as a potential therapeutic strategy in AD

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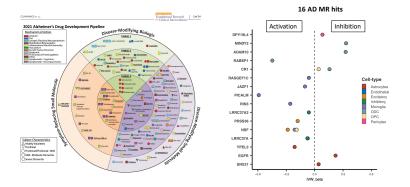
- Fine mapped in 2022 using on bulk brain tissue eQTLs (Bellenguez at al., Nat Genet, 2022) but no data on directionality or cell type
- EGFR inhibition reduces Aβ/tau pathology and reactive astrocytes in several models of AD (Lee et al., Aging Cell 2021)
- Several new BBB penetrant EGFR inhibitors (Tavassoly et al., Molecular Pharm 2021)

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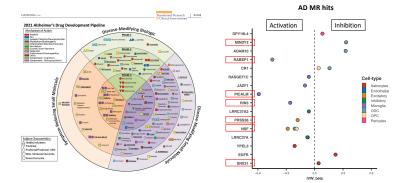
- Loss of GPNMB activity results in loss of cellular internalization of fibrillar alpha synuclein and reduced PD pathogenecity (Diaz-Ortiz et al., Science 2022)
- Levels of GPNMB in plasma correlate with PD severity (potential biomarker)
- Suggests GPNMB inhibition as a potential therapeutic strategy in PD

Focus on Alzheimer's disease: Drug development pipeline



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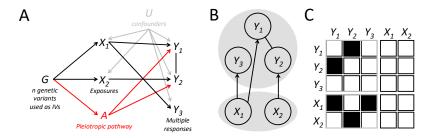
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Conclusion and Outlook

Conclusion

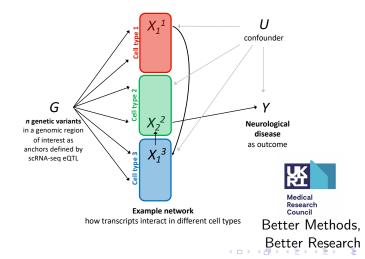
- Principled framework for single-cell Mendelian randomization target discovery in human brain phenotypes
- All samples were non-diseased control human brain samples
- Genetic evidence for 118 genes across 23 human brain diseases and behavioral traits
- 21 genes contribute to > 2 brain phenotypic outcomes (shared therapeutic strategies)
- Approach provides information on the direction of the relationship to inform therapeutic approach
- Causal mechanism and cell-types of action can inform more tailored pre-clinical experiments that may translate better to human disease and biomarker discovery

Outlook: Multi-response MR (MR²) to identify shared and distinct causes of disease



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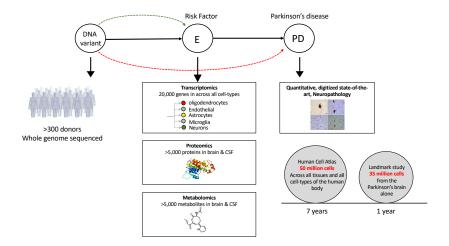
Outlook: Causal network of gene-expression across different cell-types



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Single-cell Mendelian randomisation on human brain disease and behaviour Conclusion and Outlook

Outlook: LANDMARK



Acknowledgments

- Mike Johnson
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- Dheeraj Malhotra



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Thanks to all the patients and their families who so generously donated their brains after death.

Single-cell Mendelian randomisation on human brain disease and behaviour Conclusion and Outlook



New Results

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Single-cell Mendelian randomisation identifies cell-type specific genetic effects on human brain disease and behaviour

 Alexander Haglund, ¹⁰ Verena Zuber, Yléi Yang, Maya Abouzeid, Rahel Feleke, Jeong Hun Ko, Alexi Nott, Ann C. Babtie, James D. Nills, Louwal Muhammed, Liisi Laniste, Dorgle O. Gveric, Daniel Clode, Susanna Pani, Ravishankara Bellampalli, Alyma Somani, ¹⁰ Karima McDade, Jasper J. Anink, Lucia Mesarosova, Eleonora Aronica, ¹⁰ Maria Thom, Sanjay M. Sisodiya, Prashant K. Srivastava, Dheeraj Malhotra, Julien Bryois, Leonardo Bottolo, Michael R., Johnson

doi: https://doi.org/10.1101/2022.11.28.517913

This article is a preprint and has not been certified by peer review [what does this mean?].





References I



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