

### **Presenter Financial Disclosure**

I do not have any relationships to report within the last 24 months with ACCME defined ineligible companies.





### **Unlabeled/Investigational Uses**

I will not be discussing unlabeled/investigational uses of medical devices or pharmaceuticals during this presentation.



### Age-dependent topic modelling of comorbidities in UK Biobank identifies disease subtypes with differential genetic risk

ASHG 2022 Xilin Jiang Harvard University/University of Cambridge

- Background: EHR system collects longitudinal patient medical history
- Methods
  - Age-dependent topic modelling of disease records infer latent comorbidities that represent the full patient diagnosis history
- Results
  - Identify distinct comorbidity profiles that are associated with different age-at-onsets.
  - Comorbidity profiles identify genetic heterogeneity between disease subtypes.
- Discussion: the role of individual-level comorbidity for identifying shared and distinct genetic pathways

### Longitudinal diagnosis history data are available in large data sets

#### EHR data provides individual-level age-at-onset information



Individual ID: 4427213

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**Age-at-onset analysis** 

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Multinomial 0.20 0.20 0.20 0.20 0.20 0.20 0.15 0.15 0.15 0.15 0.15 0.10. 0.10 0.10 0.10 0.10 0.05 .05 0.05 0.05 0.05 0.00 .00 0.00 0.00 0.00 70 30 40 50 60 70 80 30 40 50 60 70 80 30 40 50 60 70 80 30 40 50 60 80 30 40 50 60 70 80 30 Multinomial probability 0.20 0.20 0.20 0.20 0.20 Topics 0.15 ).15 0.15 0.15 0.15 0.10 .10 0.10 0.10 0.10 0.05 0.05 0.05 0.05 0.05 0.00 0.00 00 0.00 40 50 60 70 80 70 80 50 60 70 60 60 70 80 70 30 40 40 50 40 50 40 60 80 40 30 50 30 80 30 30 30 Multinomial probability 0.20 0.20 0.20 0.20 0.20 0.15 0.15 0.15 0.15 0.15 0.10 0.10 0.10 0.10 0.10 0.05 0.05 0.05 0.05 0.05 0.00 0.00 0.00 0.00 0.00 70 80 40 60 50 60 70 40 60 70 40 60 70 60 70 30 50 30 40 80 50 80 50 40 50 40 30 30 80 30 80 30 Age Age Age Age Age

Diseases

40 50

50 60

60

70

80

50

Age

60 70

70 80

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# Application to UK Biobank identifies 10 comorbidity trajectories

#### 1.7 million records across 348 diseases in the UK Biobank



### Early onset and late reporting of the same disease are associated with distinct comorbidities

and respiratory diseases **Essential hypertension** Asthma Essential hypertension 0.15 0.20 Type 2 diabetes Multinomial probability Multinomial probability Hypercholesterolemia 0.15 Major depressive disorder Coronary atherosclerosis 0.10 Other chronic ischemic heart disease, unspecified Angina pectoris 0.10 -**Myocardial infarction** Chronic airway obstruction Alcohol-related disorders Tobacco use disorder Atrial fibrillation and flutter 0.05 0.05 0.00 0.00 70 50 50 30 40 60 80 30 40 60 70 80 Age Age

CVD: circulatory system diseases

Topic loading examples inferred from the UK Biobank.

CER: circulatory system, endocrine/metabolic,



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- Comorbidity profiles are inferred using only diagnosis data, not genetic data.
- Genetic data could be utilised to verify the disease subtypes.

## PRS are associated with comorbidity weights





## Disease PRS stratification power varies across disease subtypes



**CVD:** Cardiovascular disease **MGND:** Male genitourinary

#### CVD MGND CER FGND



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## Genetic heterogeneity across subtypes using *Fst* and Genetic correlations



P-values for *Fst* analysis of diseases stratified by topic liabilities. 29 diseases have excessive *Fst* across subtypes identified by Comorbidity.

• Comorbidity profiles could be considered as a context for patient health status.

### Genetic risk could be modulated by comorbidity context



## SNPs have varying effect across quartiles of comorbidity weights

rs1063192 x CVD for Type 2 diabetes chr9: 9p21.3 (nearest gene:CDKN2B) P = 3.534e-07 for interaction



We identified 43 SNP x topic interactions at FDR<0.1

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### Individual-level comorbidity for identifying shared and distinct genetic pathways

1. Individuals with same disease code have distinct comorbidities.

2. The comorbidity heterogeneity among the cases imply the distinct disease subtypes.

3. The individual disease context captured by comorbidity could modulate the genetic risk effect sizes.

**MedRxiv:** Age-dependent topic modelling of comorbidities in UK Biobank identifies disease subtypes with differential genetic risk

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#### Thank you!

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