# Heritability of Human Structural Connectomes

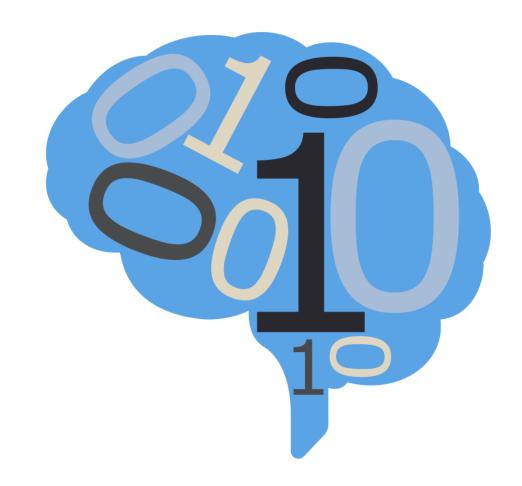
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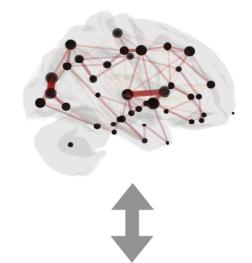
# What is heritability?

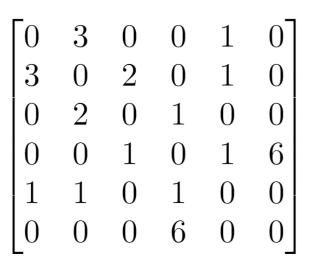
- Variations in phenotype caused by variations in genotype.
- Potentially discover relationships between diseases and genetics.

#### Are the brain connectivity patterns heritable?

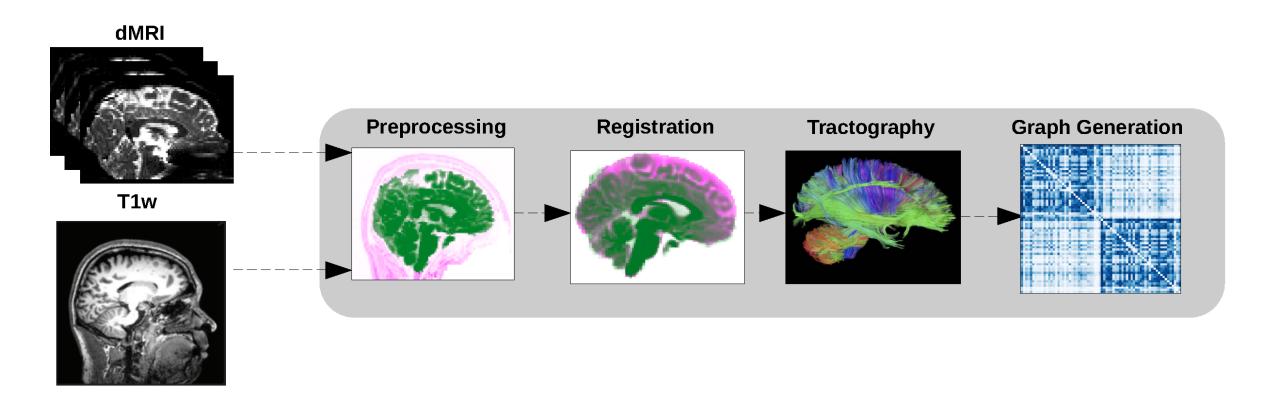
# **Brain connectivity as connectomes**

- Vertex: region of interest
- Edges: connectivity measure between a pair of vertices
- Diffusion MRI: # of estimated neuronal fibers
- Undirected: neurons have no direction



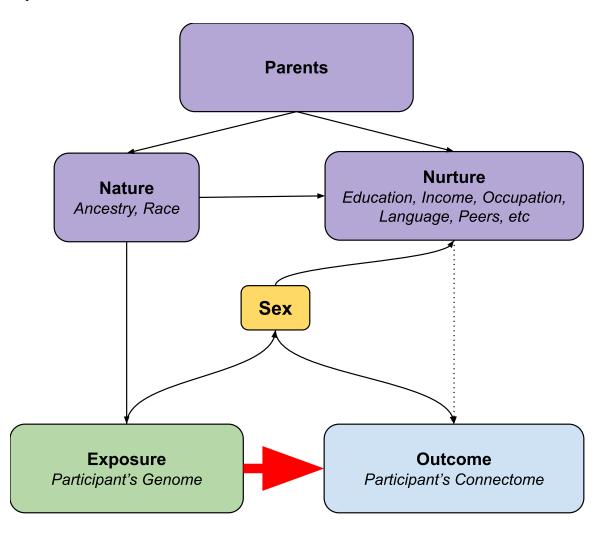


# How do we get structural connectomes?



# Heritability as causal problem

• Directed acyclic graph



## Do genomes affect connectomes?

Hypothesis:

```
H_0: F(	ext{Connectome}|	ext{Genome}) = F(	ext{Connectome})
H_A: F(	ext{Connectome}|	ext{Genome}) 
eq F(	ext{Connectome})
```

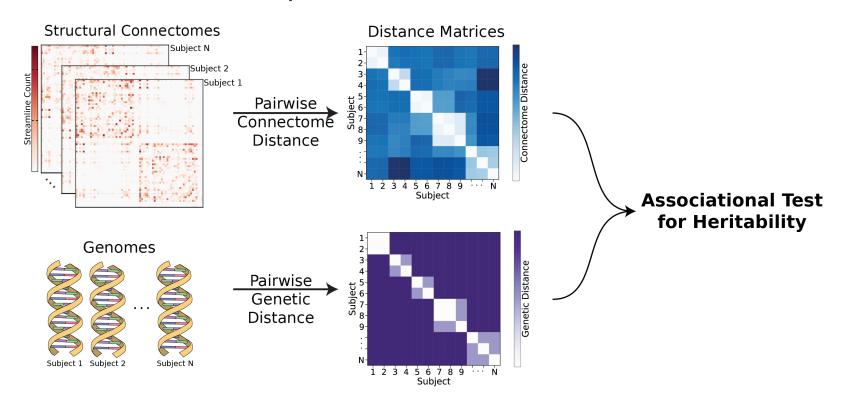
Alternatively:

```
H_0: F(	ext{Connectome}, 	ext{Genome}) = F(	ext{Connectome})F(	ext{Genome})
H_A: F(	ext{Connectome}, 	ext{Genome}) 
eq F(	ext{Connectome})F(	ext{Genome})
```

- Known as independence testing
- Test statistic: distance correlation (dcorr)
- Implication if true: there exists an associational heritability.

#### What is distance correlation?

- Measures dependence between two multivariate quantities.
  - For example: connectomes, genomes.
- Can detect nonlinear associations.
- Measures correlation between pairwise distances.



# How to compare genomes?

- Typical twin studies do not sequence genomes.
- Coefficient of kinship  $(\phi_{ij})$ 
  - Probabilities of finding particular genes as identical among subjects.
- $d(Genome_i, Genome_j) = 1 2\phi_{ij}$ .

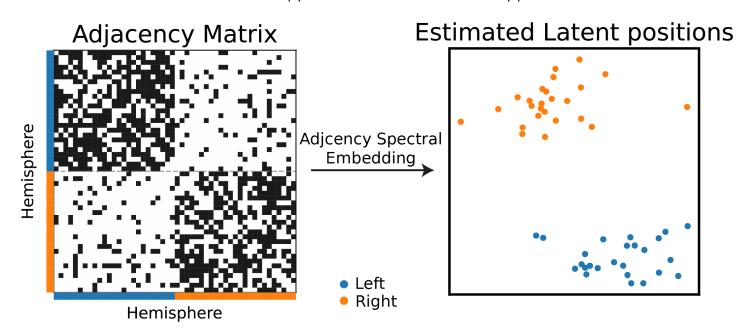
Relationship	$\phi_{ij}$	$1-2\phi_{ij}$
Monozygotic	$\frac{1}{2}$	0
Dizygotic	$\frac{1}{4}$	$\frac{1}{2}$
Non-twin siblings	$\frac{1}{4}$	$\frac{1}{2}$
Unrelated	0	1

## How to compare connectomes?

- Random dot product graph (RDPG)
  - $\circ$  Each vertex (region of interest) has a low d dimensional latent vector.

$$\circ$$
  $P[i 
ightarrow j]$  =  $\langle x_i, x_j 
angle$ 

- Latent vectors =
- $\mathsf{d}(\mathsf{Connectome}_k, \mathsf{Connectome}_l) = ||X^{(k)} X^{(l)}R||_F$

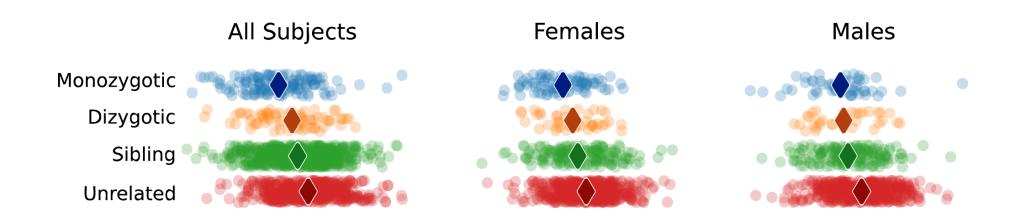


# **Human Connectome Project**

- Brain scans from identical (monozygotic), fraternal (dizygotic), non-twin siblings.
- Regions defined using Glasser parcellation

Zygosit y	Monozygoti c	Dizygotic	Non-twin siblings
N	322	212	490
Sex	196 F, 126 M	125 F, 87 M	237 F, 253 M
Age (mean )	29.6 (3.3)	28.9 (3.4)	28.3 (3.9)
Age (range )	22-36	22-36	22-37

## Genome and connectomes are dependent



Sex	All	Females	Males
p-value			

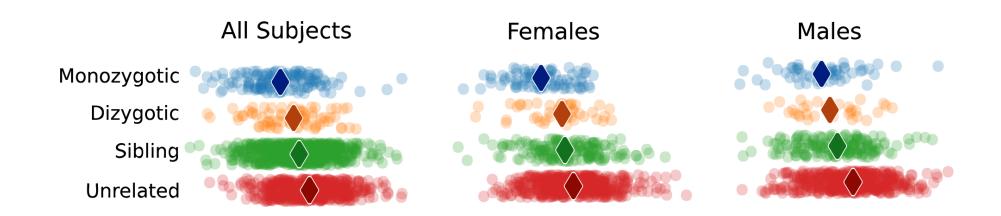
# Neuroanatomy (effect mediator)

- Literature show neuroanatomy (e.g. brain volume) is highly heritable.
- Want to test:

```
H_0: F(\text{Neuroanatomy}, \text{Genome}) = F(\text{Neuroanatomy})F(\text{Genome})
H_A: F(\text{Neuroanatomy}, \text{Genome}) \neq F(\text{Neuroanatomy})F(\text{Genome})
```

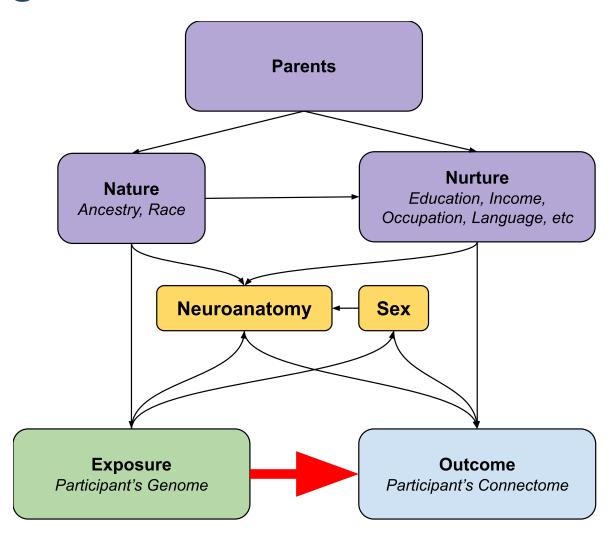
• Implication if true: causal model should include neuroanatomy.

# Genome and neuroanatomy are dependent



Sex	All	Females	Males
p-value			

# DAG including interactions of neuroanatomy



# Do genomes affect connectomes given neuroanatomy?

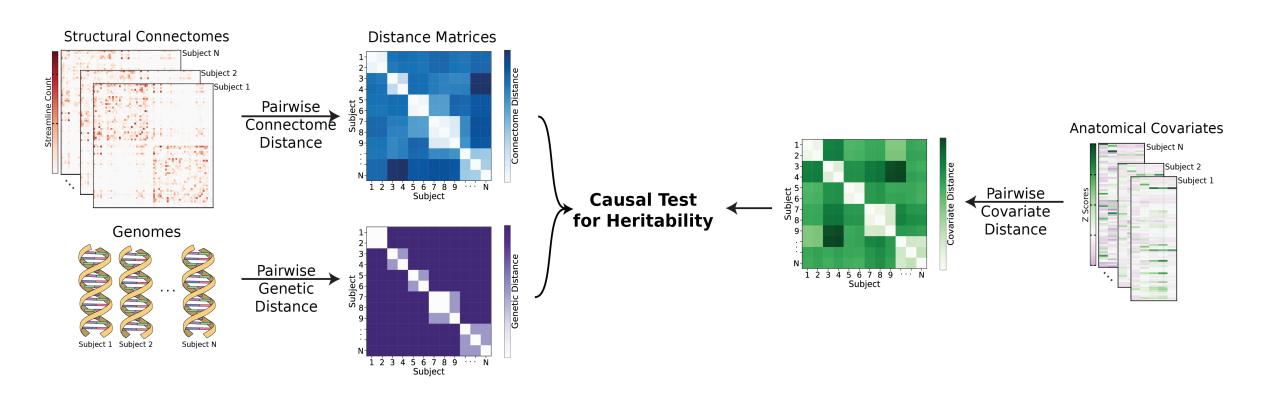
Want a conditional independence test!

```
H_0: F({\tt Conn.}, {\tt Genome|Neuro.}) = F({\tt Conn.}|{\tt Neuro.})F({\tt Genome|Neuro.})
H_A: F({\tt Conn.}, {\tt Genome|Neuro.}) 
eq F({\tt Conn.}|{\tt Neuro.})F({\tt Genome|Neuro.})
```

- Test statistic: Conditional distance correlation (cdcorr)
- Implication if true: there exists causal dependence of connectomes on genomes.

#### What is conditional distance correlation?

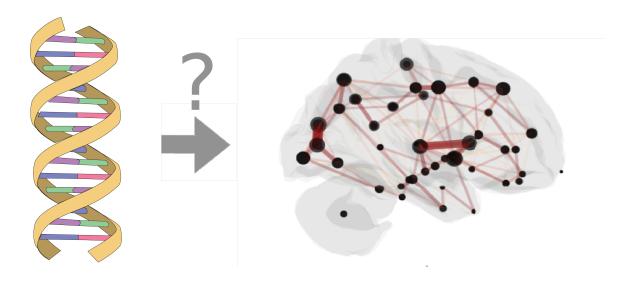
- Augment distance correlation procedure with third distance matrix.
- $d(Neuroanatomy_i, Neuroanatomy_j) = ||Neuroanatomy_i Neuroanatomy_j||_F$



# Connectomes are still dependent on genome

Sex	All	Females	Males
p-value			

# Summary



- Present a causal model for heritability of connectomes.
- Leveraged recent advances:
  - i. Statistical models for networks, allowing meaningful comparison of connectomes.
  - ii. Distance and conditional distance correlation as test statistic for causal analysis  $^1$ .
- Connectomes are dependent on genome, suggesting heritability.

<sup>&</sup>lt;sup>1</sup> Bridgeford, Eric W., et al. "Batch Effects are Causal Effects: Applications in Human Connectomics." (2021).

# **Acknowledgements**

#### **Team**



Mike Powell



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Joshua Vogelstein

## **Additional slides**

#### Causal model

- ullet X denote exposure, Y denote outcome, W denote measured covariates, Z denote unmeasured covariates
- ullet Want to estimate the effect of different exposures on the outcome, which is quantified using the backdoor formula if W and Z close all backdoor paths.

$$f_{w,z}(y|x) = \int_{\mathcal{W} imes\mathcal{Z}} f(y|x,w,z) f(w,z) \mathrm{d}(w,z)$$

Above integrates over all measured and unmeasured covariates.

$$f(y|x) = \int_{\mathcal{W} imes\mathcal{Z}} f(y|x,w,z) f(w,z|x)(w,z)$$

 Averages the true outcome distribution over the conditional distribution of the measured and unmeasured covariates.

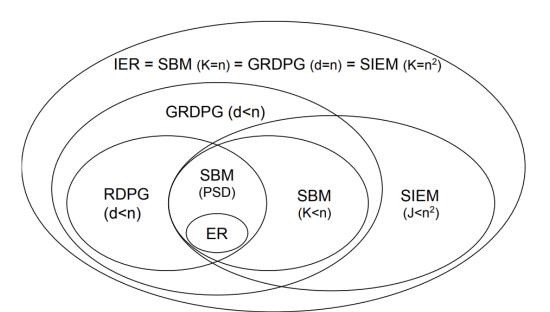
# Causal model (cont.)

- ullet We observe the triples  $(x_i,y_i,w_i)$  for  $i\in [n]$  .
- ullet Only be able to estimate the functions of (X,Y,W)
- The corresponding hypothesis test is:

$$H_0: f(y|x,w) = f(y|w) \quad ext{vs} \quad H_A: f(y|x,w) 
eq f(y|w).$$

# **Shortcomings - Network model**

- Problems with connectome estimation.
  - o Inability to determine the precise origin/termination of connections in the cortex.
    - -> false negatives
  - Crossing fibers
    - -> false positives
- RDPG can only represent subset of independent edge networks.



# **Shortcomings - Model assumptions**

- No interaction between genome and environment
- No epistatsis
  - Effect of one gene is dependent on another
  - Ex: black hair and baldness
- No dominance effects
- Strong assumptions in genetic distances

#### What are environmental effects?

- Shared
  - Common experiences of siblings living in the same household.
    - household income, the family's living situation, the dynamics between the parents, food consumed
- Non-shared
  - Everything else
  - Epigenetics
  - Luck
  - schools, peers

## Random dot product graphs

- Adjacency spectral embedding
- representation of the vertices of the graphs into d dimensions via its singular value decomposition, given by  $A = USU^{\top}$  where  $U \in \mathbb{R}^{n \times n}$  is the orthogonal matrix of eigenvectors and  $S \in \mathbb{R}^{n \times n}$  is a diagonal matrix containing the eigenvalues of A ordered by magnitude.
- $ASE(A) = \hat{X} = \hat{U}\hat{S}^{1/2}$  where  $\hat{U} \in \mathbb{R}^{n \times d}$  contains the first d columns of U, which correspond to the largest eigenvectors, and  $\hat{S} \in \mathbb{R}^{d \times d}$  is the submatrix of \$ S \$ corresponding to the d largest eigenvalues in magnitude.