

Trait simulation

Huwenbo Shi

August 8, 2016

1 Types of simulations

1.1 GWAS simulation

1.1.1 Single quantitative trait simulation

- Users specify
 1. Genotype matrix
 2. Effect size vector for SNPs
 3. Desired heritability
 4. Desired missing rate
- Model: $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$
- \mathbf{y} : trait values for n individuals
- \mathbf{X} : the $n \times p$ genotype matrix, provided by the user (allelic? standardized?)
- $\boldsymbol{\beta}$: the p -vector of effect sizes, provided by the user
- $\boldsymbol{\epsilon}$: the n -vector of environmental effect, drawn by the program such that the desired amount of heritability is achieved.

1.1.2 Single case-control trait simulation

- Users specify
 1. Genotype matrix
 2. Effect size vector for SNPs
 3. Desired heritability
 4. Prevalence of the cases
 5. Desired number of cases and controls
- Simulation procedure
 1. First, simulate liability based on $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$
 2. Then, select liability threshold based on prevalence
 3. Finall, draw cases and controls according to the specified numbers

1.2 Two correlated traits

- Users specify
 1. Genotype matrix
 2. SNP effect size vectors for 2 traits

3. Desired heritability for 2 traits
 4. Desired missing rate for 2 traits (or prevalence and number of cases and controls for case-control traits)
 5. Desired phenotypic correlations between the two traits
- Model: $\mathbf{y}_1 = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$, $\mathbf{y}_2 = \mathbf{X}\boldsymbol{\gamma} + \boldsymbol{\delta}$
 - Simulation procedure
 1. First, simulate the genetic component $\mathbf{X}\boldsymbol{\beta}$ and $\mathbf{X}\boldsymbol{\gamma}$
 2. Compute $\text{Var}[\mathbf{X}\boldsymbol{\beta}]$, $\text{Var}[\mathbf{X}\boldsymbol{\gamma}]$, and $\text{Cov}[\mathbf{X}\boldsymbol{\beta}, \mathbf{X}\boldsymbol{\gamma}]$
 3. Find σ_{ϵ}^2 , σ_{δ}^2 , and $\sigma_{\epsilon\delta}$ and draw $\boldsymbol{\epsilon}$ and $\boldsymbol{\delta}$, from bivariate normal (parameterized by σ_{ϵ}^2 , σ_{δ}^2 , and $\sigma_{\epsilon\delta}$), such that the desired amount of heritability and phenotypic correlation are achieved.
 4. Dichotomize liability for case-control traits

1.3 Variance component simulation

1.3.1 Two variance components (σ_g^2 , σ_e^2)

- Users specify
 1. Heritability of the trait (σ_g^2)
 2. Genotype matrix
 3. Missing rate
 4. Mean effect
- Model $\mathbf{y} = \boldsymbol{\mu} + \mathbf{X}\mathbf{u} + \boldsymbol{\epsilon}$
- \mathbf{y} : n -vector of trait values
- \mathbf{X} : Standardized $n \times p$ genotype matrix
- \mathbf{u} : p -vector of SNP effect ($\mathbf{u} \sim N(0, \sigma_g^2/p)$)
- $\boldsymbol{\mu}$: mean effect (can be a function of genotypes, i.e. $\boldsymbol{\mu}_i = \tilde{\boldsymbol{\mu}}_i + \mathbf{X}_{ik}?$)
- Simulate from $N(\boldsymbol{\mu}, \sigma_g^2 \mathbf{X} \mathbf{X}^\top / n + \sigma_e^2 \mathbf{I})$

1.3.2 Multiple variance components (σ_g^2 , σ_i^2 , σ_e^2)

- Users specify
 1. σ_g^2 , and all σ_i^2
 2. Genotype matrix
 3. All other design matrices for each variance component
 4. Missing rate
 5. Mean effect
- Model $\mathbf{y} = \boldsymbol{\mu} + \mathbf{X}\mathbf{u} + \sum_i \mathbf{W}_i \mathbf{v}_i + \boldsymbol{\epsilon}$
- \mathbf{X} : standardized $n \times p$ genotype matrix
- \mathbf{W}_i : design matrices for each variance component

1.4 General multiple correlated traits and multiple variance components simulation

- Users specify

1. Mean effects $\boldsymbol{\mu}_i$ for m traits
 2. Genotype matrix \mathbf{X}
 3. Design matrices for each variance component \mathbf{W}_i
 4. Heritability: σ_{gj}^2 (heritability of trait j)
 5. Other variance components: σ_{ij}^2
 6. Genetic covariances: σ_{gjk} for the pair of traits \mathbf{y}_j and \mathbf{y}_l
 7. Random noise covariances: σ_{ejk} for the pair of traits \mathbf{y}_j and \mathbf{y}_l
 8. Other covariances: σ_{ijk} for the pair of traits \mathbf{y}_j and \mathbf{y}_l
- Simulate from the model

$$N \left(\begin{bmatrix} \boldsymbol{\mu}_i \\ \vdots \\ \boldsymbol{\mu}_m \end{bmatrix}, \mathbf{C}_g \otimes \frac{\mathbf{X} \mathbf{X}^\top}{n} + \sum_i \mathbf{C}_i \otimes \frac{\mathbf{W}_i \mathbf{W}_i^\top}{n} + \mathbf{C}_e \otimes \mathbf{I} \right), \text{ where } \mathbf{C}_* = \begin{bmatrix} \sigma_{*1}^2 & \cdots & \sigma_{*1m} \\ \vdots & \ddots & \vdots \\ \sigma_{*m1} & \cdots & \sigma_{*m}^2 \end{bmatrix}$$