

Rare-Variant Association Testing for Sequencing Data with the Sequence Kernel Association Test (Michael C. Wu et al., 2011)

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Presentation 1: Introduction, Motivation and Overview

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Outline

Challenge in rare-variant association test

Limitations for previous methods

- Burden test

- C-alpha test

Overview of Sequence Kernel Association Test (SKAT)

- Flexible

- Computationally efficient

Bonus example

Background Knowledge

- Minor Allele Frequency (MAF)

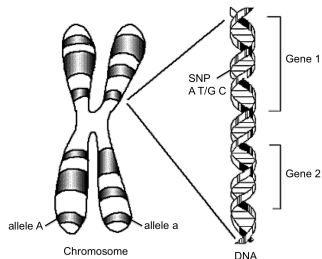
Frequency at which least common allele occurs in a given population

- Rare variant

Variant with a MAF less than 1-5% (in SKAT: 3%)

- Genome Wide Association Study (GWAS)

An examination of many **common** genetic variants in different individuals to see if any variant is associated with a trait



Challenge in rare-variant association test

Subject	V1	V2	V3	Disease
1	1	1	0	1
2	0	0	1	1
3	0	0	0	0
4	0	0	0	0
5	0	0	0	0

- Multiple rare variants within the same functional unit e.g. exon of a gene — unexplained genetic component of complex traits (Missing heritability)
- Traditional association test of common variants: underpowered unless sample sizes or effect sizes are very large
- Need to collectively analyze instead of individually

Limitations for previous methods

- Burden test: summarize/collapse rare variants in a region

$l_{ij} = 0, 1 \text{ or } 2$ – number of minor alleles at variant j for individual i

r_i – number of variants that carry at least one copy of the minor allele

n_i – total number of rare variants, $r_i = \sum_{j=1}^{n_i} 1(l_{ij} > 0)$

Count-Based Proportion: $E(Y_i) = \beta_0 + \lambda \frac{r_i}{n_i} + \beta X_i$

Dichotomize (Cohort Allelic Sum Test, CAST):

$E(Y_i) = \beta_0 + \lambda 1(r_i > 0) + \beta X_i$

Weighted Sum Test (WST): $\text{logit } P(Y_i = 1) = \beta_0 + r_i^* + \beta X_i + \epsilon_i$,

where $r_i^* = \sum_{j=1}^{n_i} \frac{l_{ij}}{w_j}$

- Limitation: assume that all rare variants influence the phenotype in the same direction and with the same magnitude

Limitations for previous methods

- C-alpha test:

Compares the expected variance to the actual variance of the distribution of allele frequencies

- Limitation:

Only apply to case-control data

Cannot adjust for covariate (population stratification)

Need permutation sometimes (computationally expensive)

Overview of Sequence Kernel Association Test (SKAT)

- Model: $\text{logit } P(Y_i = 1) = \beta_0 + \sum_{j=1}^p b_j G_{ij} + \beta X_i$

or $E(Y_i) = \beta_0 + \sum_{j=1}^p b_j G_{ij} + \beta X_i$

i: individual, $i = 1, \dots, n$

j: p genetic variants within a functional region – common and rare

$X_i = (X_{i1}, X_{i2}, \dots, X_{im})$: covariates, e.g. age, gender

Y_i : phenotype (dichotomous or continuous)

$G_i = (G_{i1}, G_{i2}, \dots, G_{ip})$: genotype, $G_{ij} = 0, 1$ or 2

- Assumption: $b_j \sim (0, (w_j \sigma)^2)$

$$H_0 : \mathbf{b} = 0 \Leftrightarrow H_0 : \sigma^2 = 0 \text{ (variance-component test)}$$

- Allows for different directions and magnitudes of genetic effects

Overview of Sequence Kernel Association Test (SKAT)

- Variance-component score statistic:

$$Q = (y - \hat{y}_0)' K (y - \hat{y}_0), \hat{y}_0: \text{fitted value under } H_0$$
$$= \|WG'(y - \hat{y}_0)\|^2$$

kernel $K = GWWG'$, weight $W = \text{diag}(w_j)$

$$= \sum_{j=1}^p w_j^2 \|G_j(\mathbf{y} - \hat{\mathbf{y}}_0)\|^2$$
$$= \sum_{j=1}^p w_j^2 \sum_{i=1}^n [G_{ij}(y_i - \hat{y}_0)]^2$$
$$\sim \text{mixture of } \chi_1^2$$

Weighted sum of individual score statistic $\mathbf{S}_j = G_j(\mathbf{y} - \hat{\mathbf{y}}_0)$

- Only requires fitting the null model

Overview of Sequence Kernel Association Test (SKAT)

- SKAT: weighted sum of score statistic

$$\text{logit } P(Y_i = 1) = \beta_0 + \sum_{j=1}^p \mathbf{b}_j G_{ij} + \beta X_i$$

$$Q_j = \sum_{j=1}^p w_j^2 \sum_{i=1}^n [G_{ij}(y_i - \hat{y}_0)]^2$$

- Burden: weighted sum of genetic variants:

$$\text{logit } P(Y_i = 1) = \beta_0 + \sum_{j=1}^p \mathbf{w}_j \beta_{\mathbf{B}} G_{ij} + \beta X_i$$

$$Q = \left[\sum_{i=1}^n (y_i - \hat{y}_0) \left(\sum_{j=1}^p w_j G_{ij} \right) \right]^2$$

Overview of Sequence Kernel Association Test (SKAT)

- Choice of weight: w_j

$$\text{Beta}((MAF_j); 1, 25)$$

Upweight rare-variant
downweight common-variant

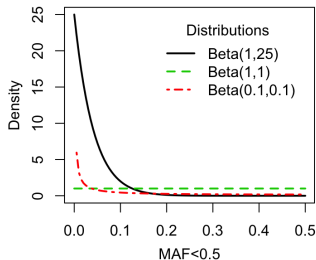
$$\text{Beta}((MAF_j); 1, 1)$$

Uniform[0, 1]

$$\text{Beta}((MAF_j); 0.1, 0.1)$$

steeper than $\text{Beta}(1, 25)$ at (0, 0.01)

Comparison of Beta Distributions



Overview of Sequence Kernel Association Test (SKAT)

- Choice of kernel function: $(K(G_i, G_{i'}))_{n \times n}$ positive semidefinite

Weighted linear kernel function $K(G_i, G_{i'}) = \sum_{j=1}^P w_j^2 G_{ij} G_{i'j}$

Linear genetic effects

Weighted quadratic kernel $K(G_i, G_{i'}) = (1 + \sum_{j=1}^P w_j G_{ij} G_{i'j})^2$

Both linear and quadratic genetic effects

$\text{logit } P(Y_i = 1) = \beta_0 + f(G_i) + \beta X_i, H_0 : f(G) = 0$

Weighted IBS kernel $K(G_i, G_{i'}) = \sum_{j=1}^P w_j \text{IBS}(G_{ij}, G_{i'j})$

Identity by state (IBS), number of alleles that share IBS

Free of assumption of additivity, allows for interaction between variants

Sequence Kernel Association Test (SKAT)

- Test for association between phenotype and a collection of rare and common variants in sequencing-based association studies
- Robust to direction and magnitude
- Allow for covariate adjustment
Works for both continuous and dichotomous phenotype
- Only need to fit null model
No permutation needed for p-value
- $f(G)$: allow for epistatistic effects (interaction between genetic variants); family data
Y: regression based, easily extended to survival, longitudinal and multivariate phenotypes

Next Steps

- Estimation of power and sample size
- Simulations; compare to previous methods
- Application to Dallas Heart Study Data? (still waiting for response)
- Math in appendix A

Thank you!

References

- Wu, M. et. al. Rare-variant association testing for sequencing data with the sequence kernel association test. *American Journal of Human Genetics*, 92: 841-853, 2013.
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