

KMgene: a unified R package for gene-based association analysis for complex traits

Qi Yan¹, Zhou Fang² and Wei Chen^{1,2}

UPMC CHILDREN'S

¹Division of Pulmonary Medicine, Allergy and Immunology, Department of Pediatrics, Children's Hospital of Pittsburgh of UPMC, University of Pittsburgh, ²Department of Biostatistics, Graduate School of Public Health, University of Pittsburgh

Background

- One widely used gene-based test is the sequence kernel association test (SKAT, Wu et al., 2011) based on a KM regression framework.
- After SKAT was introduced for testing independent samples with continuous and binary traits, a number of methods and corresponding tools have been developed to extend the approach to complex traits.
- We introduce KMgene, which combines SKAT-type methods for complex traits and extends them to include their corresponding optimal tests. KMgene can perform association tests between a set of genetic variants and familial, multivariate, longitudinal or survival traits (Table 1, Yan et al., 2018).

Table 1. A summary of funct	ions in KMgene package
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	Regular (KM)	Optimal (KM-O)	Interaction (KM-Int)
Continuous family (F-KM)	Chen et al. 2013	Extended	NA
Binary family (Fb-KM)	Yan et al. 2015	Extended	NA
Continuous multivariate (M-KM)	Maity et al. 2012	Extended	NA
Continuous multivariate family (MF-KM)	Yan et al. 2015	Extended	NA
Continuous longitudinal (L-KM)	Yan et al. 2015	Yan et al. 2015	Extended
Survival (CoxKM)	Chen et al. 2014	NA	NA

Method

KMgene works in two steps:

- The first step with function names, *prefix_Null_Model*, fits the model under the null hypothesis (i.e., the genetic effects are zero). The estimates of covariate parameters and covariance matrix are obtained at this step. The covariance matrix can account for relatedness in families, correlation between multivariate traits or between times for longitudinal data.
- The second step with function names, *prefix*, constructs the test statistic and calculates the *p*-value. We use the parameter estimates from step one to construct the test statistic. Since the parameters are estimated under the null hypothesis and used for all genes, they only need to be calculated once for the whole genome-wide analysis, which greatly reduces the computation time.
- According to our derivation, the test statistic follows a mixture of χ^2 distributions and thus we can compute the *p*-values analytically, also leading to improvement in computation.
- The KM statistics can be extended to the optimal test by combining with burden statistics. Analogously, our optimal tests consist of two steps for fitting null models (*prefixO_Null_Model*) and calculating *p*-values (prefixO).

Input:

- Genotypes pre-grouped in genes and coded as 0, 1, 2 for the number of copies of minor allele (i.e., additive genetic model);
- Traits and covariates;
- · Family pedigree when analyzing familial data.

Output:

· Gene-level p-values

Results (real data example)

Here, as an illustrative example, we apply MFKM_Null_Model() and MFKM() to carry out a gene-based genome wide association test of the correlated lung function phenotypes FEV1 (Forced Expiratory Volume in One Second) and FEV1/FVC (Forced Vital Capacity) ratio (Yan, et al., 2015). We identified COL6A6 associated with these two traits (Fig 1) and COL6A6 is known to be in the chronic obstructive pulmonary dis-ease related regions based on Rat Genome Database (RGD)

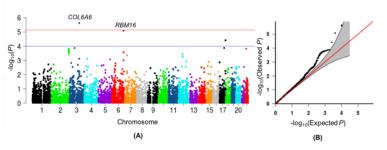


Fig1. (A) Genome wide gene-based results of MFKM on lung function data. Each dot represents p-value of a gene. (B) QQ plot of p-values from the lung function analysis, with 95% pointwise confidence band (gray area).

Conclusions

- This R package adapts GLMM to conduct gene-based tests for complex traits and uses Cox model for survival trait.
- KMgene can handle genome-wide genotypic datasets with reasonable computational time.
- KMgene currently uses the linear kernel that is the most commonly used kernel in genetic studies.

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