

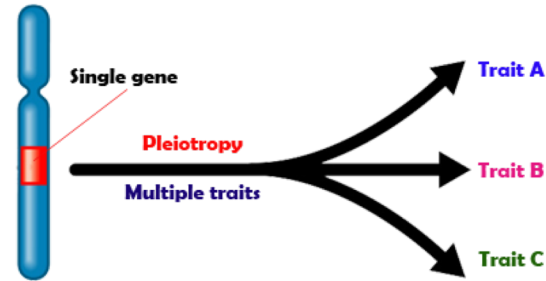
Associating Multivariate Quantitative Phenotypes with Genetic Variants in Family Samples with a Novel Kernel Machine Regression Method

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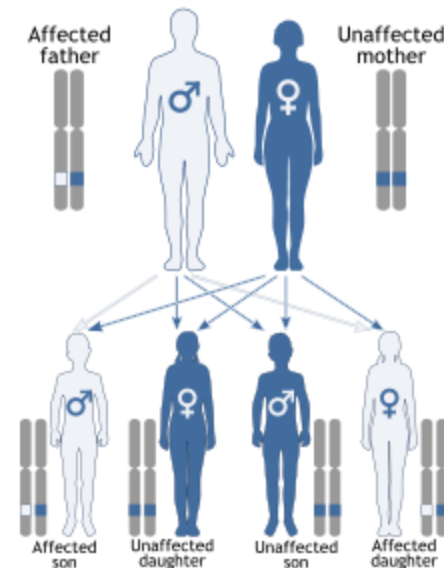
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Motivation



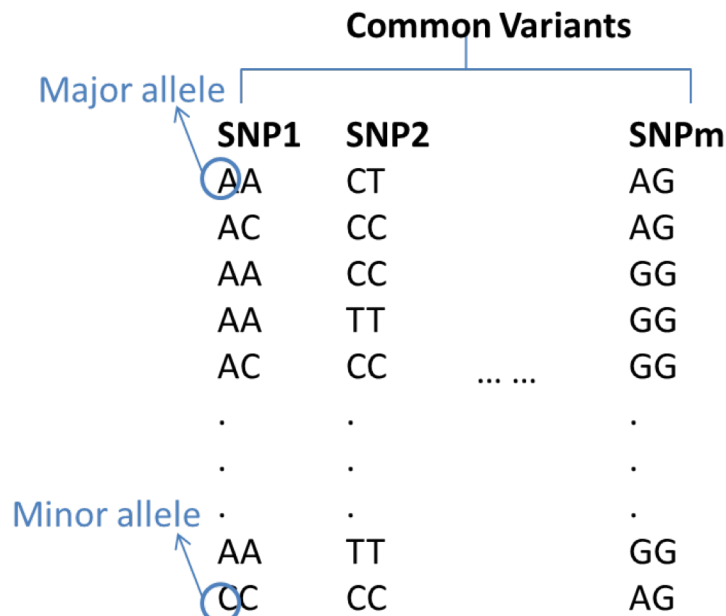
- Phenotypes:

- Genetic studies have been conducted to collect multiple correlated phenotypes for one complex disease. Jointly modeling multiple phenotypes can improve the statistical power [Sivakumaran S, et al. AJHG. 2011];
- Family based designs have been widely used [Spielman RS, et al. AJHG. 1993].
Appropriately handling familial correlation can retain Type I error rate;



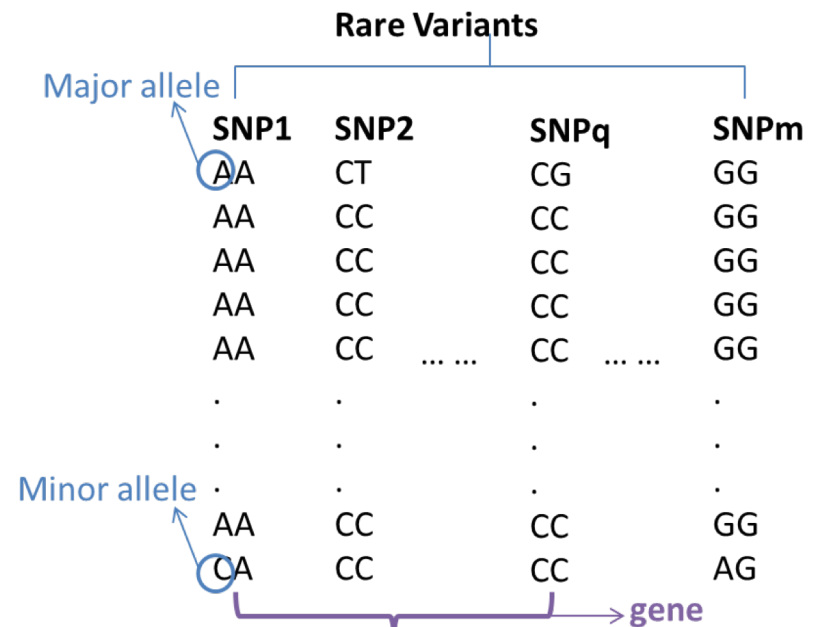
Motivation

- Genotypes:
 - Common variants (e.g. $MAF \geq 0.05$): single marker test;
 - Rare variants (e.g. $MAF < 0.05$): test at gene level (e.g. SKAT).



$MAF = (\# \text{ of minor alleles}) / 2n$

$MAF > 0.05$ (common variant)



$MAF = (\# \text{ of minor alleles}) / 2n$

$MAF < 0.05$ (rare variant)

Aims

- Association test between multiple quantitative phenotypes and genes in family samples
 - Rare variants are assigned into genes;
 - Family structure is considered;
 - Correlated quantitative phenotypes are tested simultaneously.

Methods

➤ Kernel Machine (KM) Regression for Linear Mixed Model:

Let there be n subjects with q genetic variants. The $n \times 1$ vector of the quantitative trait \mathbf{y} follows a linear mixed model:

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{G}\boldsymbol{\gamma} + \mathbf{u} + \boldsymbol{\varepsilon}$$

- \mathbf{X} is an $n \times p$ covariate matrix,
- $\boldsymbol{\beta}$ is a $p \times 1$ vector containing parameters for the fixed effects (an intercept and $p - 1$ covariates),
- \mathbf{G} is an $n \times q$ genotype matrix for the q genetic variants of interest,
- $\boldsymbol{\gamma}$ is a $q \times 1$ vector for the random effects of the q genetic variants,
- $\boldsymbol{\varepsilon}$ is an $n \times 1$ vector for the random error,
- \mathbf{u} is an $n \times 1$ vector for the random effects due to covariates (e.g., correlation between phenotypes or relatedness in families)

$$\boldsymbol{\gamma} \sim N(0, \tau \mathbf{W}) \quad H_0: \tau = 0$$

$$\mathbf{u} \sim N(0, \mathbf{K})$$

$$\boldsymbol{\varepsilon} \sim N(0, \sigma_E^2 \mathbf{I})$$

where \mathbf{W} is a predefined $q \times q$ diagonal weight matrix for each variant, and \mathbf{K} is an $n \times n$ covariance matrix

Methods

➤ Kernel Machine (KM) Regression for Linear Mixed Model:

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{G}\boldsymbol{\gamma} + \mathbf{u} + \boldsymbol{\varepsilon}$$

$$\boldsymbol{\gamma} \sim N(0, \tau\mathbf{W})$$

For the linear mixed model, the log likelihood is

$$l = C - \frac{1}{2} \log|\boldsymbol{\Sigma}| - \frac{1}{2} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \boldsymbol{\Sigma}^{-1} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})$$

$$\boldsymbol{\Sigma} = \tau\mathbf{G}\mathbf{W}\mathbf{G}' + \mathbf{K} + \sigma_E^2\mathbf{I}$$

To derive the score test for $H_0: \tau = 0$, we take the first derivative with respect to τ

Score function: $\frac{dl}{d\tau} = -\frac{1}{2} \operatorname{tr}(\underbrace{\boldsymbol{\Sigma}^{-1}\mathbf{G}\mathbf{W}\mathbf{G}'}_{\uparrow \text{Fixed}}) + \frac{1}{2} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \boldsymbol{\Sigma}^{-1} \underbrace{\mathbf{G}\mathbf{W}\mathbf{G}'\boldsymbol{\Sigma}^{-1}}_{\uparrow \text{Test statistic}} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})$

Methods

➤ Kernel Machine (KM) Regression for Linear Mixed Model:

Under the null hypothesis, the linear mixed model is $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{u} + \boldsymbol{\varepsilon}$, and the estimates are

$$\widehat{\boldsymbol{\Sigma}} = \widehat{\mathbf{K}} + \widehat{\sigma}_E^2 \mathbf{I}$$

$$\widehat{\boldsymbol{\beta}} = (\mathbf{X}'\widehat{\boldsymbol{\Sigma}}^{-1}\mathbf{X})^{-1}\mathbf{X}'\widehat{\boldsymbol{\Sigma}}^{-1}\mathbf{y}$$

Replacing the variance components with their maximum likelihood estimators (MLEs), we have

$$Q = (\mathbf{y} - \mathbf{X}\widehat{\boldsymbol{\beta}})' \widehat{\boldsymbol{\Sigma}}^{-1} \mathbf{G} \mathbf{W} \mathbf{G}' \widehat{\boldsymbol{\Sigma}}^{-1} (\mathbf{y} - \mathbf{X}\widehat{\boldsymbol{\beta}})$$

as the test statistic. Under the null hypothesis, the variance of the residual is: $Var(\mathbf{y} - \mathbf{X}\widehat{\boldsymbol{\beta}}) = \mathbf{P}_0$

The statistic Q is a quadratic form of $\mathbf{y} - \mathbf{X}\widehat{\boldsymbol{\beta}}$ and follows a mixture of chi-square distributions under H_0 . Thus,

$$Q \sim \sum_{i=1}^q \lambda_i \chi_{1,i}^2$$

where λ_i are the eigenvalues of the matrix $\mathbf{W}^{\frac{1}{2}} \mathbf{G}' \widehat{\boldsymbol{\Sigma}}^{-1} \mathbf{P}_0 \widehat{\boldsymbol{\Sigma}}^{-1} \mathbf{G} \mathbf{W}^{\frac{1}{2}}$

Methods

➤ Kernel Machine Regression for Quantitative phenotypes in Family Data (MF-KM):

Under the null hypothesis,

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{h} + \boldsymbol{\varepsilon}$$

- \mathbf{y} is a vector of quantitative trait (i.e., $\mathbf{y} = (y_{11}, y_{12}, y_{21}, y_{22}, \dots, y_{m1}, y_{m2})$ where m is the number of individuals),
- $\mathbf{X}\boldsymbol{\beta}$ is the fixed effects of covariates,
- \mathbf{h} is the random effect of correlated phenotypes corresponding to the polygenic contribution,
- $\boldsymbol{\varepsilon}$ is the random effect of correlated phenotypes corresponding to the random environmental contribution.

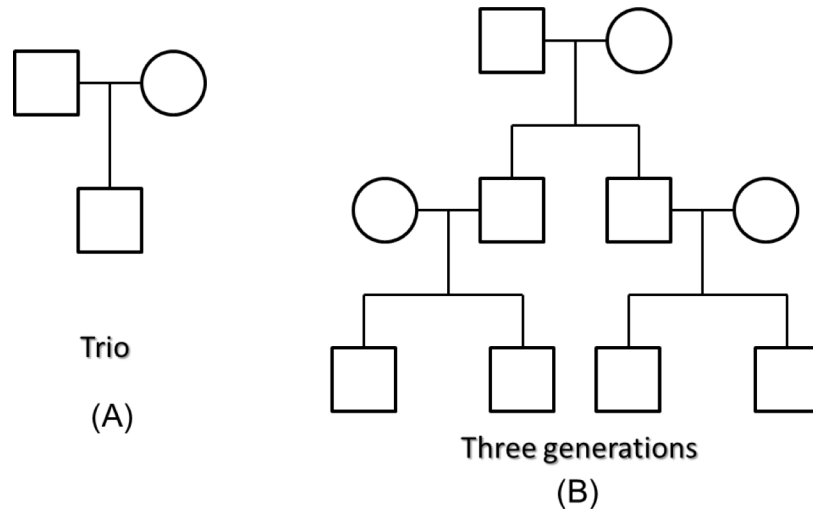
$$\text{Var}(\mathbf{y}) = \underbrace{\boldsymbol{\Phi}}_{\text{kinship}} \otimes \underbrace{\begin{pmatrix} \sigma_{G1}^2 & \sigma_{G12} \\ \sigma_{G12} & \sigma_{G2}^2 \end{pmatrix}}_{\text{polygenic variances}} + \mathbf{I} \otimes \underbrace{\begin{pmatrix} \sigma_{E1}^2 & \sigma_{E12} \\ \sigma_{E12} & \sigma_{E2}^2 \end{pmatrix}}_{\text{environmental variances}} = \boldsymbol{\Sigma}$$

Methods

➤ Simulation Studies

Genotypes:

- Trios:
 - One genotype dataset = 300 trios \times 30 rare variants;
 - Total = 100 genotype datasets (1000 sets of phenotypes for each set of genotypes).
- Three-generation families:
 - One genotype dataset = 100 families \times 30 rare variants;
 - Total = 100 genotype datasets (1000 sets of phenotypes for each set of genotypes).



Methods

➤ Simulation Studies

Phenotypes:

- Type I error rate: 1000 sets of phenotypes for each genotype dataset (independent);

$$\mathbf{y}_i = 0.05 \cdot \mathbf{X}_{1i} + 0.5 \cdot \mathbf{X}_{2i} + \mathbf{e}_i$$

$$\begin{aligned} \text{Var}(\mathbf{y}_i) &= \mathbf{\Phi}_i \otimes \begin{pmatrix} \sigma_{G1}^2 & \sigma_{G12} \\ \sigma_{G12} & \sigma_{G2}^2 \end{pmatrix} + \mathbf{I}_{3 \times 3} \otimes \begin{pmatrix} \sigma_{E1}^2 & \sigma_{E12} \\ \sigma_{E12} & \sigma_{E2}^2 \end{pmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0.5 \\ 0 & 1 & 0.5 \\ 0.5 & 0.5 & 1 \end{bmatrix} \otimes \begin{pmatrix} 1 & 0.8 \\ 0.8 & 1 \end{pmatrix} + \mathbf{I}_{3 \times 3} \otimes \begin{pmatrix} 1 & 0.8 \\ 0.8 & 1 \end{pmatrix} \end{aligned}$$

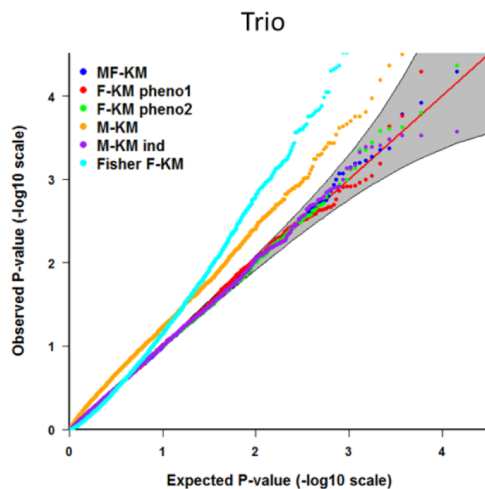
- Power: 1000 sets of phenotypes for each genotype dataset (Causal variants(+/-)
= 30%/0%; 20%/10%; 20%/0%; 13%/7%).

$$\mathbf{y}_i = 0.05\mathbf{X}_{1i} + 0.5\mathbf{X}_{2i} + \boldsymbol{\beta}_1\mathbf{G}_1 + \boldsymbol{\beta}_2\mathbf{G}_2 + \dots + \boldsymbol{\beta}_k\mathbf{G}_k + \mathbf{e}_i$$

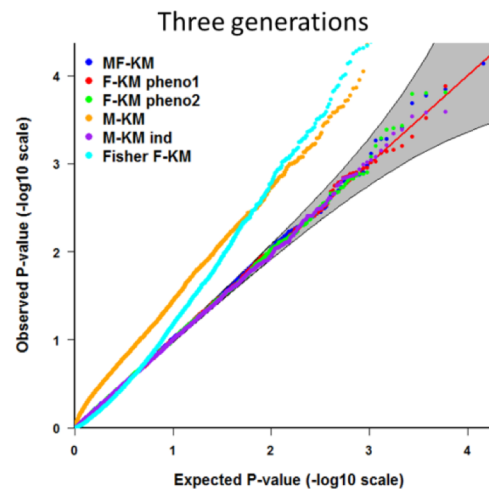
Results

➤ Simulation of the Type I Error Rate:

		$\alpha=0.05$	$\alpha=0.01$	$\alpha=0.005$	$\alpha=0.001$
Trios	MF-KM	0.0497	0.0108	0.0051	0.0014
	F-KM pheno1	0.0511	0.0113	0.0057	0.0007
	F-KM pheno2	0.0473	0.0103	0.0051	0.0012
	M-KM	0.0861	0.0211	0.0125	0.0031
	M-KM ind	0.0497	0.0108	0.0047	0.0011
	Fisher F-KM	0.0796	0.0285	0.0192	0.0072
Three generations	MF-KM	0.0503	0.0105	0.0049	0.0010
	F-KM pheno1	0.0519	0.0104	0.0049	0.0010
	F-KM pheno2	0.0496	0.0104	0.0051	0.0010
	M-KM	0.1270	0.0384	0.0222	0.0062
	M-KM ind	0.0495	0.0094	0.0051	0.0011
	Fisher F-KM	0.0830	0.0292	0.0200	0.0078



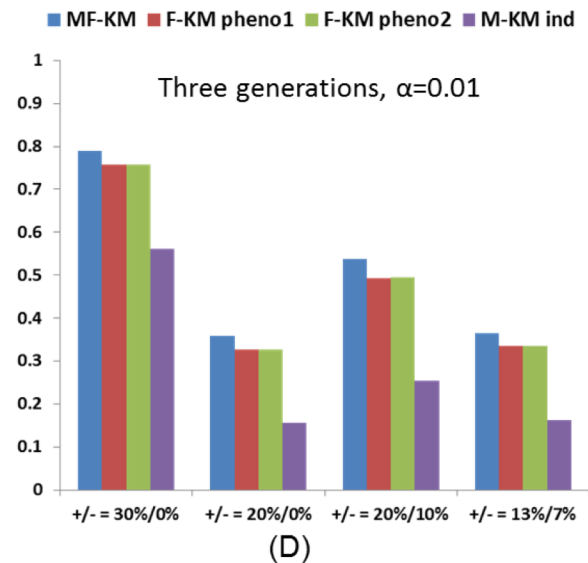
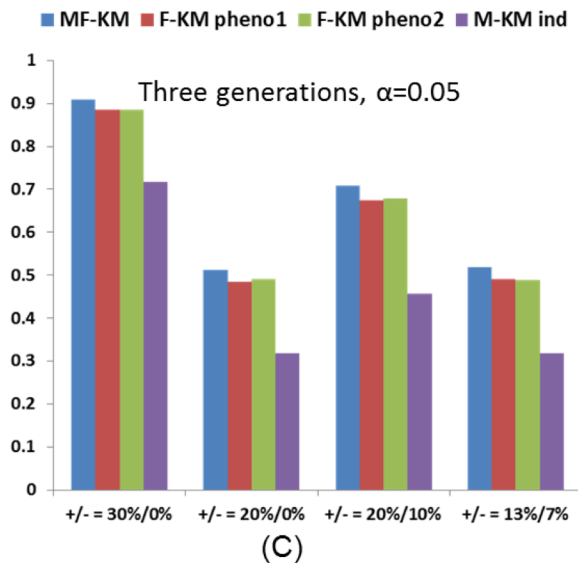
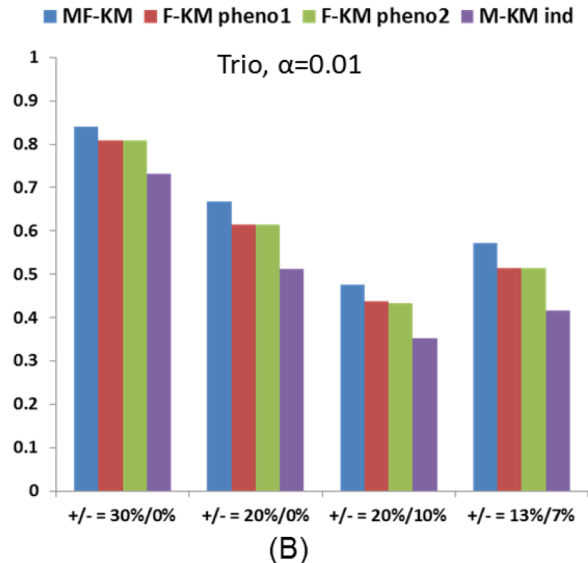
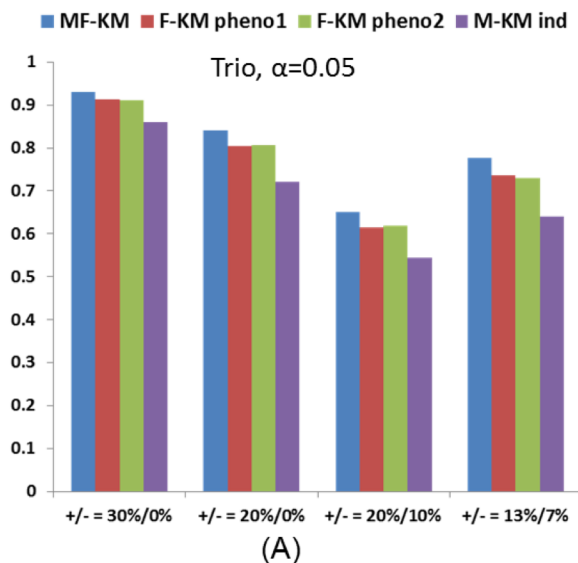
(A)



(B)

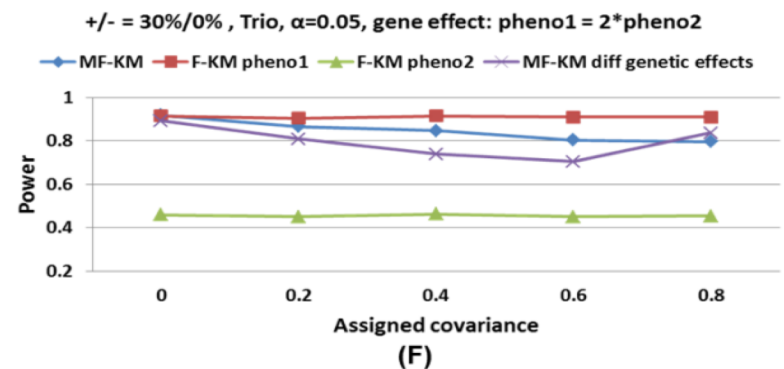
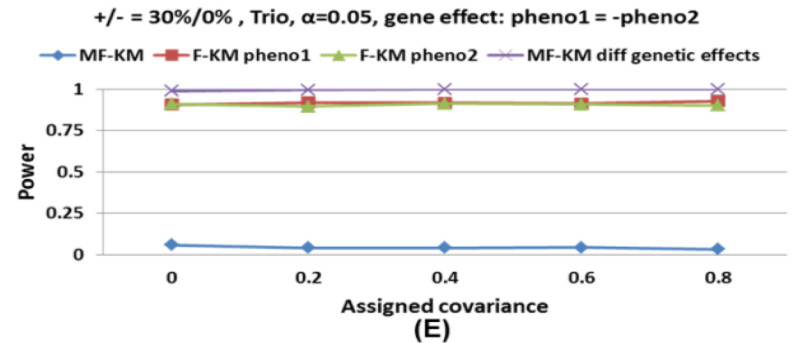
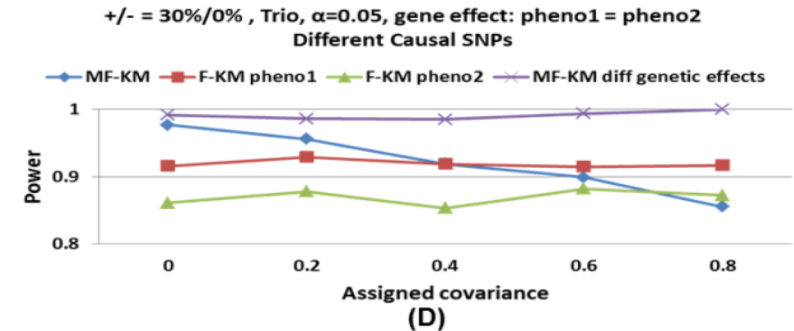
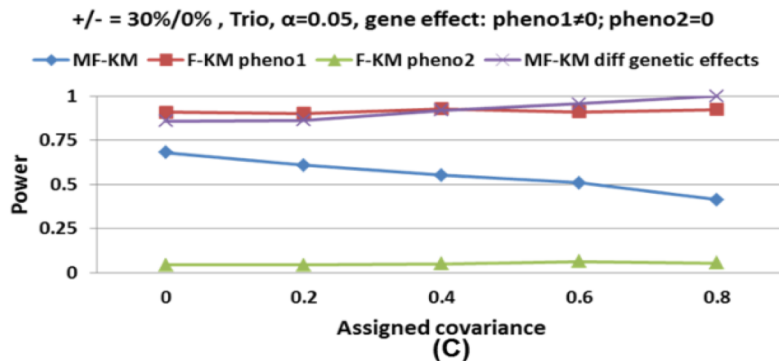
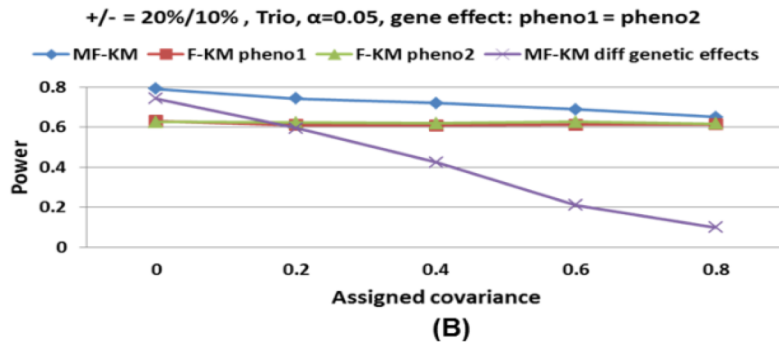
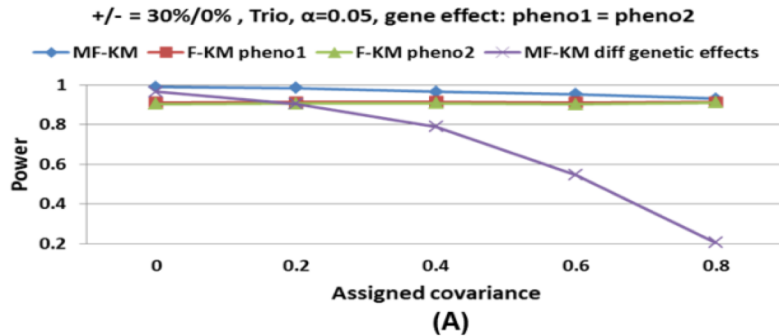
Results

➤ Statistical Power Comparison:



Results

➤ Statistical Power Comparison:

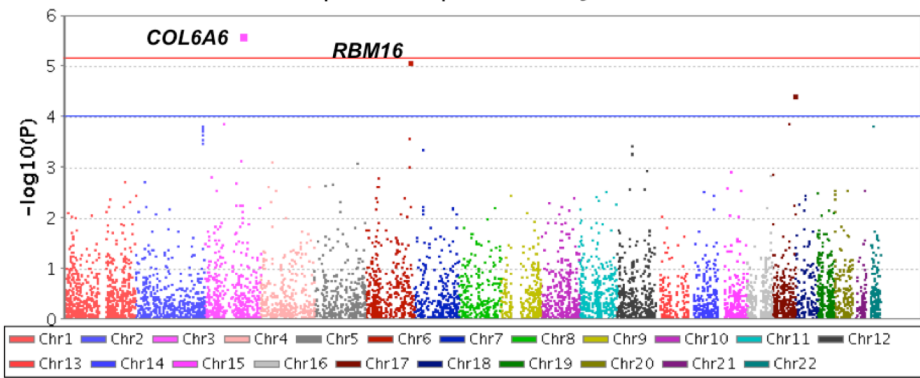


Results

➤ Analysis of Genome Wide Lung Function Data:

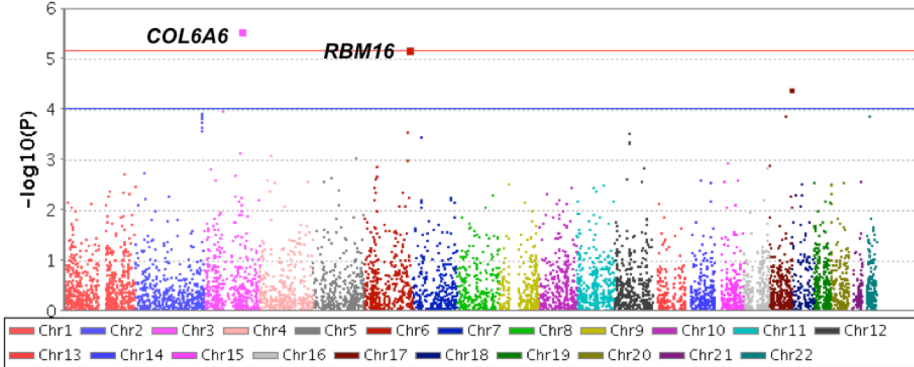
- 579 subjects, including 316 samples from 13 families;
- 658,502 SNPs were genotyped, where 67,121 are rare variants (MAF<0.05);
- Assigned rare variants to a gene if they are located within a 5kb flank;
- 7,064 genes were used in the analysis;
- Carried out gene-based genome wide association tests of the correlated lung function phenotypes FEV1 (Forced Expiratory Volume in One Second) and FEV1/FVC (Forced Vital Capacity) ratio using MF-KM adjusted for age, gender and height.

FEV₁ and FEV₁/FVC Jointly from MF-KM



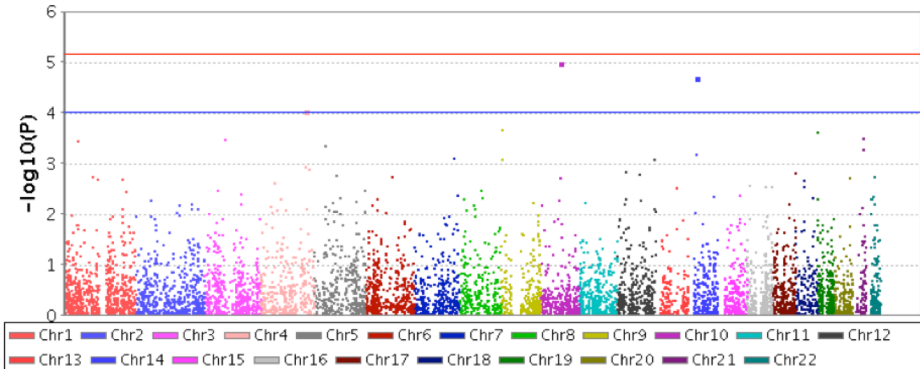
(A)

FEV₁ from F-KM



(B)

FEV₁/FVC from F-KM



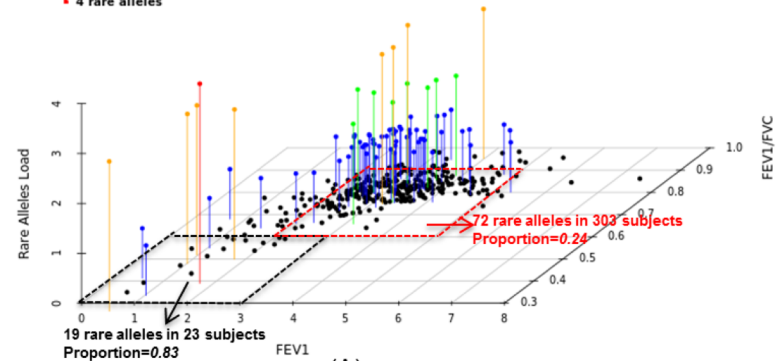
(C)

Two genes



Rare Alleles Load across 7 Rare Variants in COL6A6

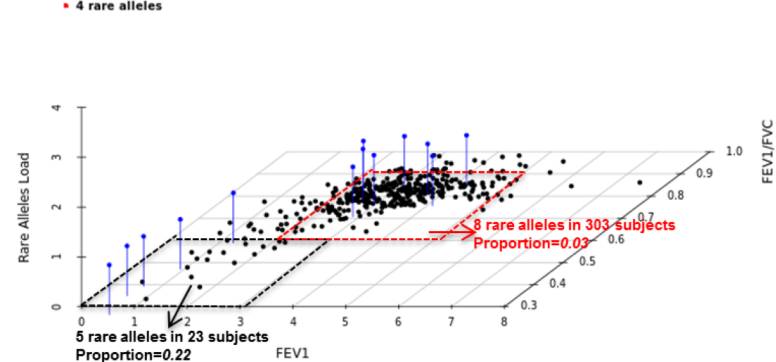
- 0 rare alleles
- 1 rare alleles
- 2 rare alleles
- 3 rare alleles
- 4 rare alleles



(A)

Rare Alleles Load across 2 Rare Variants in RBM16

- 0 rare alleles
- 1 rare alleles
- 2 rare alleles
- 3 rare alleles
- 4 rare alleles



(B)

Summary

- Developed the MF-KM statistic using a linear mixed model framework to analyze multivariate data with quantitative traits in family-based studies.
- MF-KM retains the correct Type I error rate, and achieves the best power performance.
- The software is available (<http://www.pitt.edu/~qiy17/Softwares.html>).

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