

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

**Qi Yan**

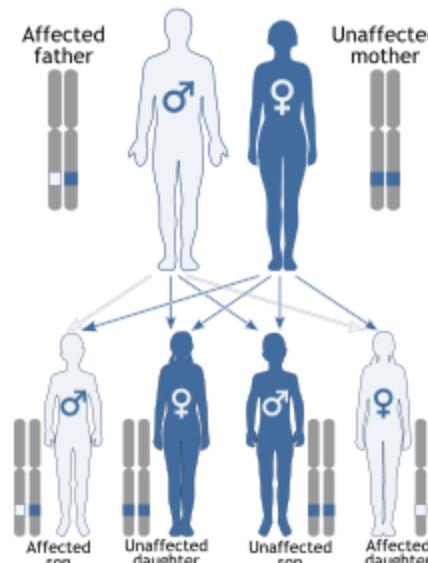
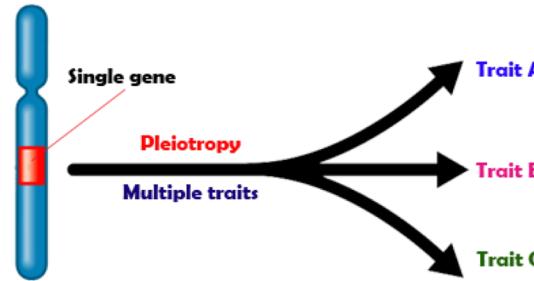
Department of Pediatrics, University of Pittsburgh  
Children's Hospital of Pittsburgh of UPMC

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# Motivation

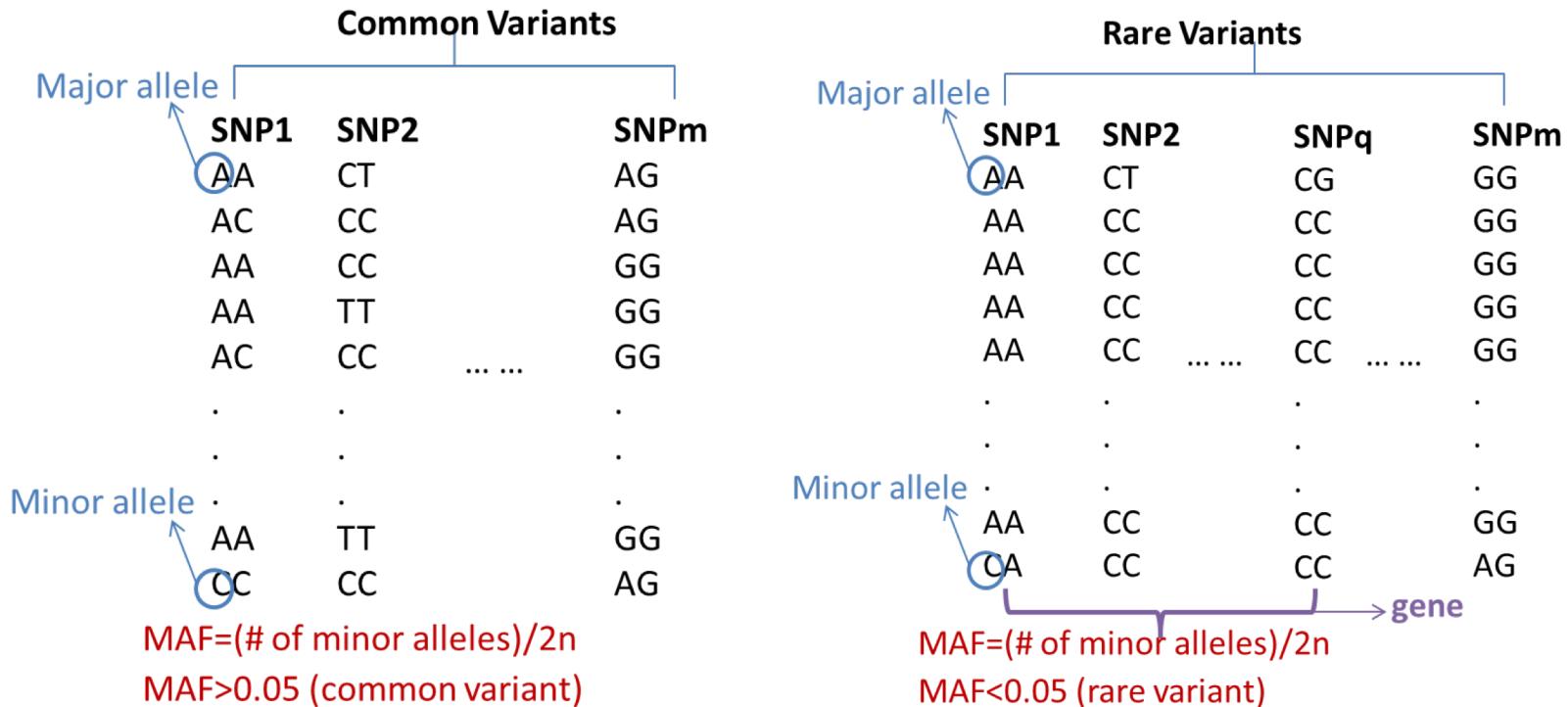
- Phenotypes:

- 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).



# 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

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

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

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

**Score function:**  $\frac{dl}{d\tau} = -\frac{1}{2} \text{tr}(\Sigma^{-1} \mathbf{G} \mathbf{W} \mathbf{G}') + \frac{1}{2} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})' \Sigma^{-1} \mathbf{G} \mathbf{W} \mathbf{G}' \Sigma^{-1} (\mathbf{y} - \mathbf{X}\boldsymbol{\beta})$

↑    ↑  
Fixed    Test statistic

# 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

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

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

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  $\lambda_i$  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.

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

↑  
kinship      ↑  
polygenic variances      ↑  
environmental variances

# Methods

## ➤ Simulation Studies

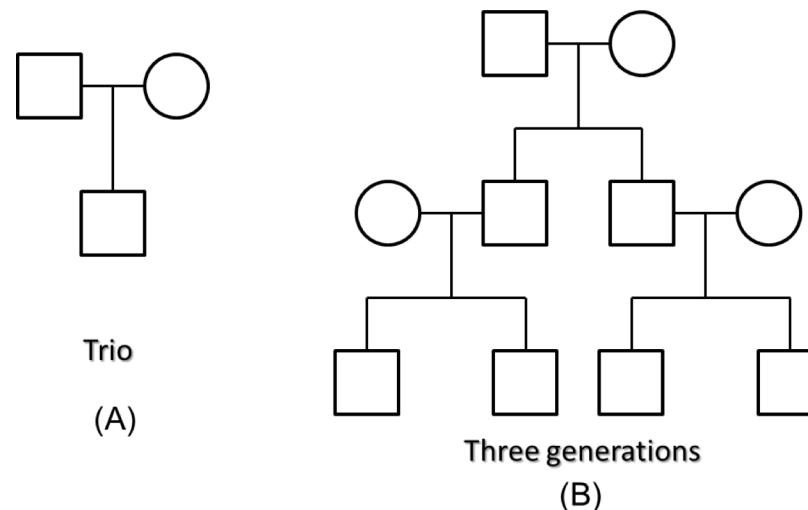
### Genotypes:

- 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);

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

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

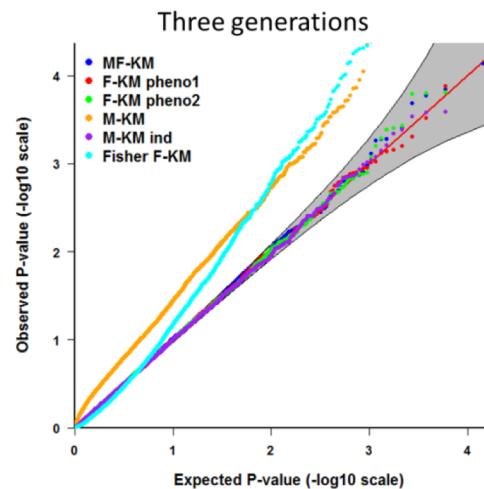
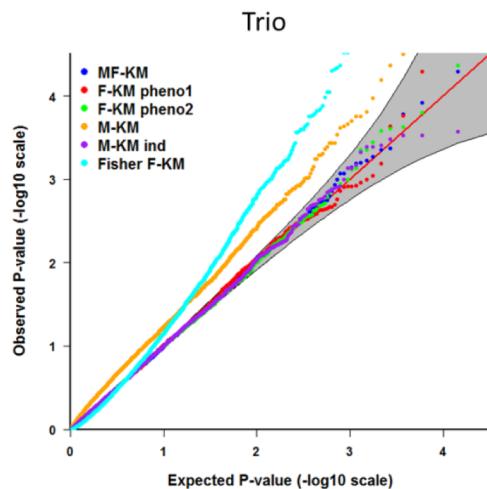
- Power: 1000 sets of phenotypes for each genotype dataset (Causal variants(+/-) = 30%/0%; 20%/10%; 20%/0%; 13%/7%).

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

# Results

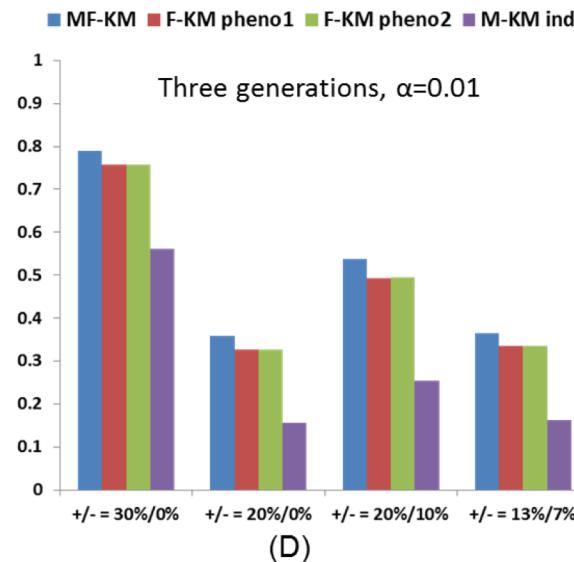
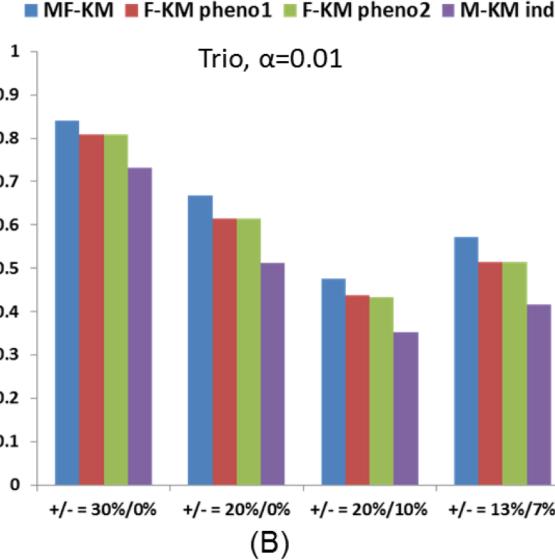
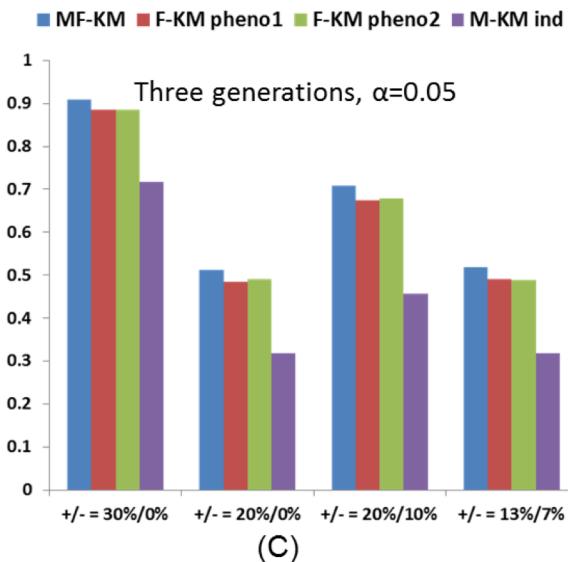
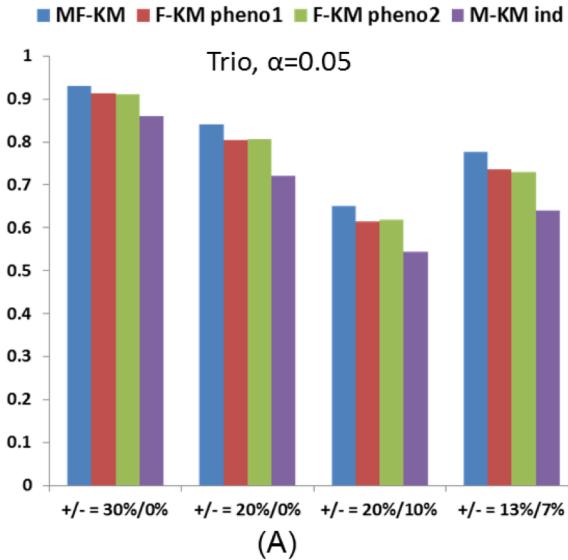
## ➤ Simulation of the Type I Error Rate:

	$\alpha=0.05$	$\alpha=0.01$	$\alpha=0.005$	$\alpha=0.001$
Trios	<b>MF-KM</b>	0.0497	0.0108	0.0051
	<b>F-KM pheno1</b>	0.0511	0.0113	0.0057
	<b>F-KM pheno2</b>	0.0473	0.0103	0.0051
	<b>M-KM</b>	<b>0.0861</b>	<b>0.0211</b>	<b>0.0125</b>
	<b>M-KM ind</b>	0.0497	0.0108	0.0047
	<b>Fisher F-KM</b>	<b>0.0796</b>	<b>0.0285</b>	<b>0.0192</b>
Three generations	<b>MF-KM</b>	0.0503	0.0105	0.0049
	<b>F-KM pheno1</b>	0.0519	0.0104	0.0049
	<b>F-KM pheno2</b>	0.0496	0.0104	0.0051
	<b>M-KM</b>	<b>0.1270</b>	<b>0.0384</b>	<b>0.0222</b>
	<b>M-KM ind</b>	0.0495	0.0094	0.0051
	<b>Fisher F-KM</b>	<b>0.0830</b>	<b>0.0292</b>	<b>0.0200</b>



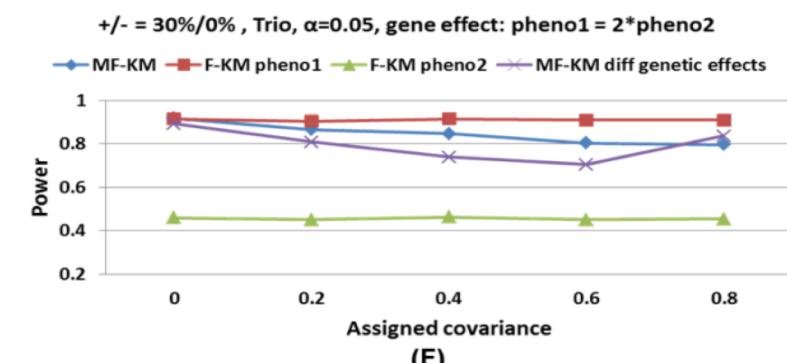
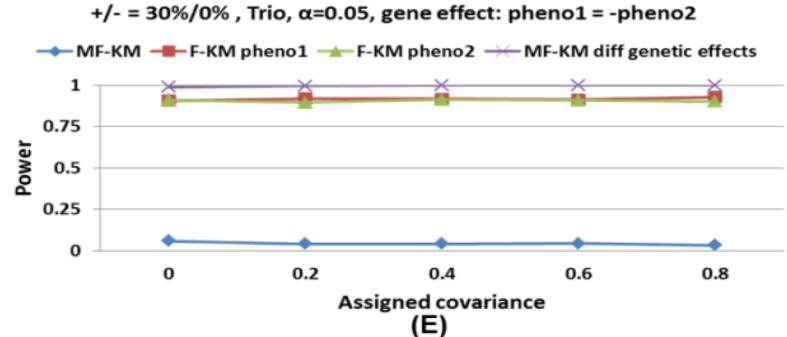
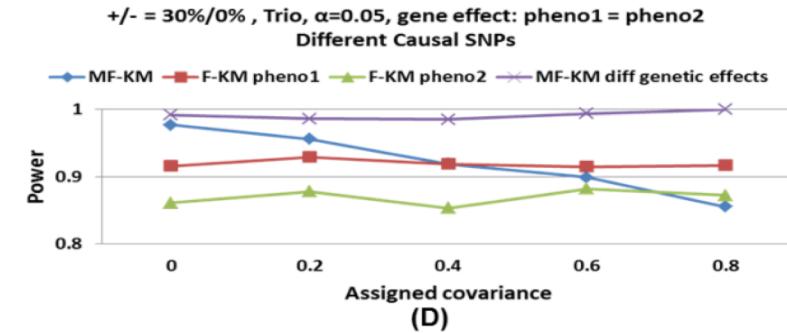
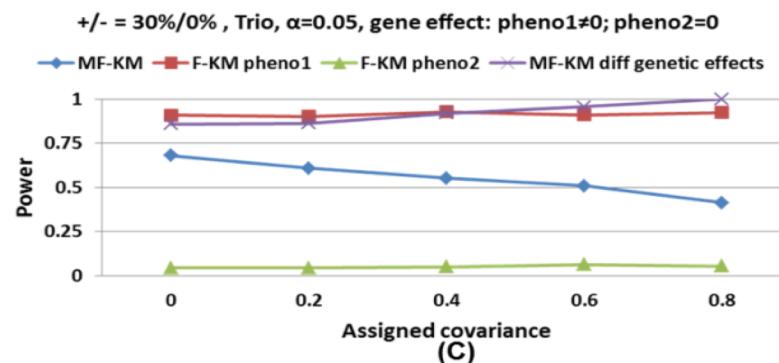
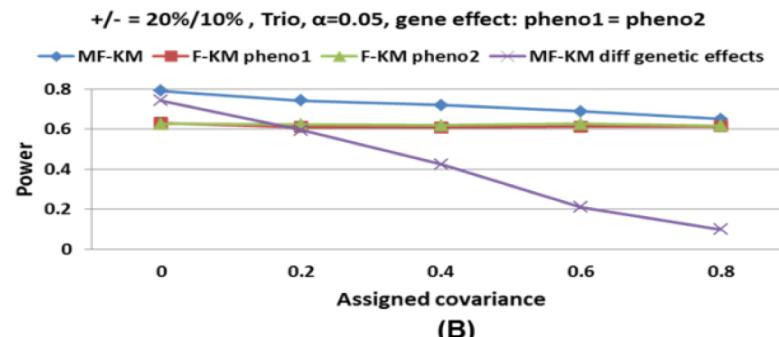
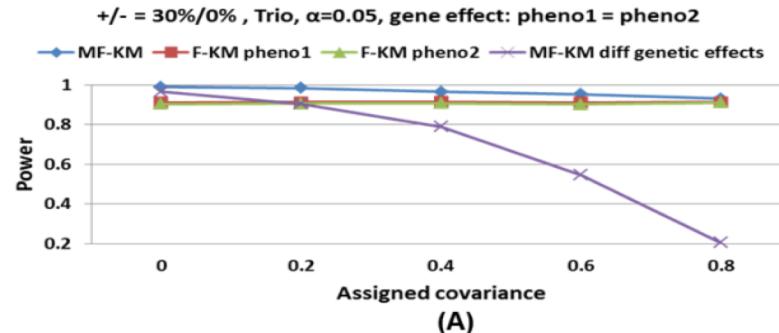
# Results

## ➤ Statistical Power Comparison:



# Results

## ➤ Statistical Power Comparison:

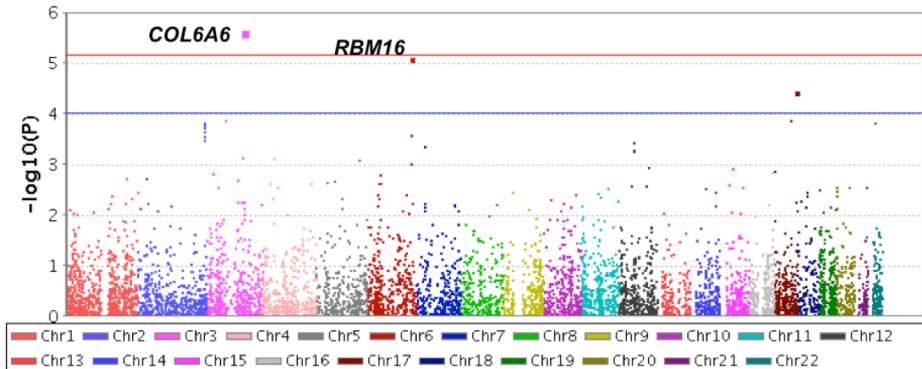


# Results

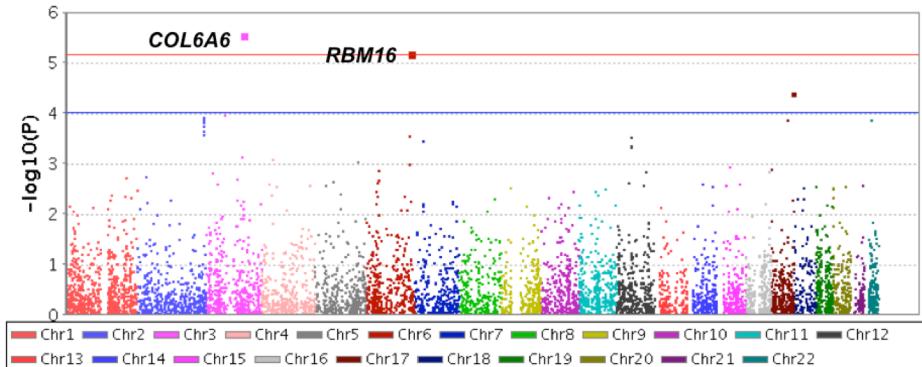
## ➤ Analysis of Genome Wide Lung Function Data:

- 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.

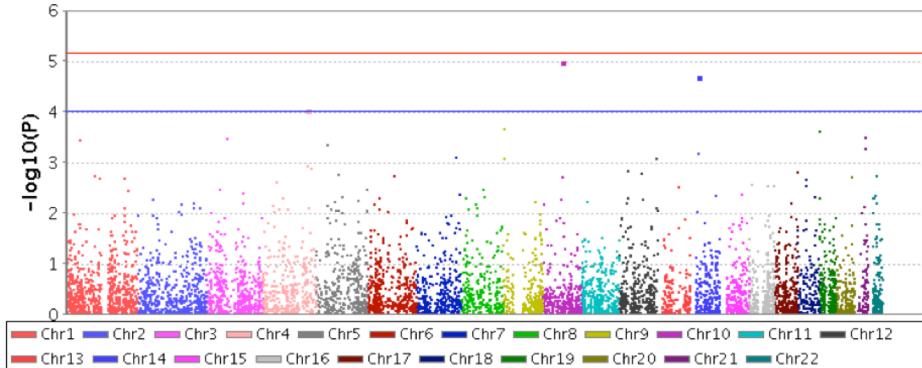
### FEV<sub>1</sub> and FEV<sub>1</sub>/FVC Jointly from MF-KM



FEV<sub>1</sub> from F-KM



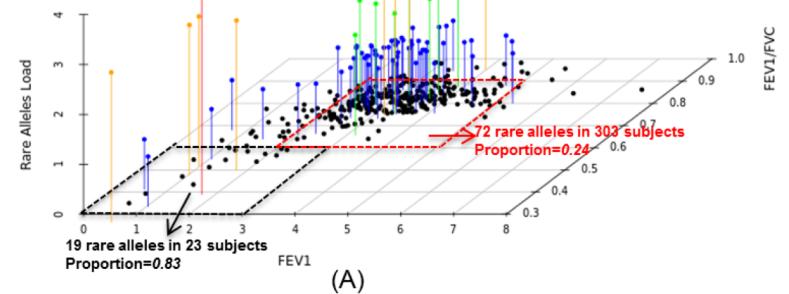
FEV<sub>1</sub>/FVC from F-KM



Two genes

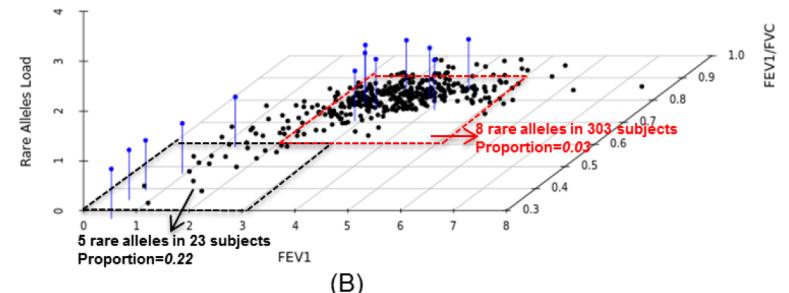
### Rare Alleles Load across 7 Rare Variants in COL6A6

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



### Rare Alleles Load across 2 Rare Variants in RBM16

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



# 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>).

# Acknowledgements

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