

Genome-Wide Study of miRNA Regulated Gene Expressions Networks in Association with Emphysematous Lung Destruction

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COPD

Chronic Obstructive Pulmonary Disease (COPD)

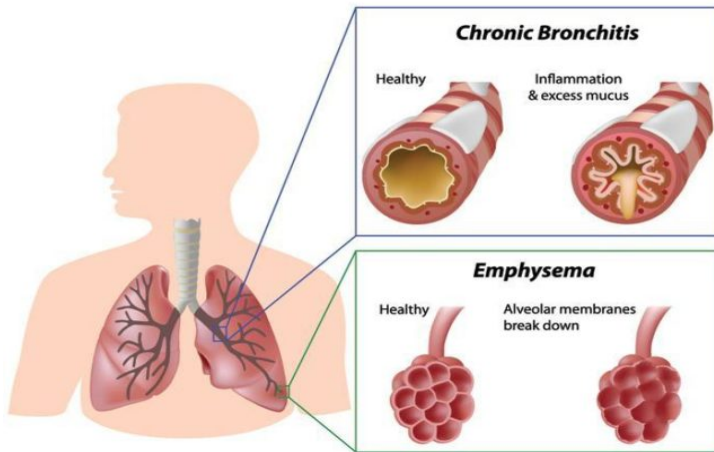


Figure 1: COPD is a progressive fatal lung disease with the potential for major complications and is often eventually fatal.

COPD

- COPD is the third leading cause of death in U.S, but the precise mechanism of the development of COPD has not fully understood yet.
- Recent studies show that several microRNAs are deregulated in COPD patients and microRNA signature becomes altered based on the severity of COPD.
- microRNA is associated with various human diseases such as cardiovascular disease, idiopathic pulmonary fibrosis (IPF) and Alzheimer Disease.

Research Goal:

- whether the progression of severe emphysema from no emphysema alters microRNA signature.
- the relationship of altered microRNAs and their target mRNA changes in emphysema severity within an individual lung.

COPD Data Preprocessing

- six subjects has undergone lung transplantation for sever COPD and two subjects are donors without COPD.
- Each subject has up to 12 slices of lung from apex to bottom.
- For each slice of lung, the emphysema severity, 397 microRNA expression profiles, and 22011 mRNA expression profiles are measured.
- The measurements for some slices are not available.
- The data is available through GEO using GSE27597 and GSE49881.

COPD Data

Response: $y = \text{emphysema severity}(\log(Lm))$

Demographic: $x_1 = \text{COPD}$, $x_2 = \text{packyears}$, $x_3 = \text{age}$
 $x_4 = \text{sex}$, $x_5 = \text{slices}$

miRNA: $\text{miRNA}_j, j = 1, 2, \dots, 397$

mRNA: $\text{mRNA}_k, k = 1, 2, \dots, 22011$

Where

packyears = The number of packs of cigarettes consumed each year.

slice = A multi-level factor variable ranging from 2 to 13, referring to the position where it locates in lung from apex to base.

Nonlinearity between age, packyears and lLm

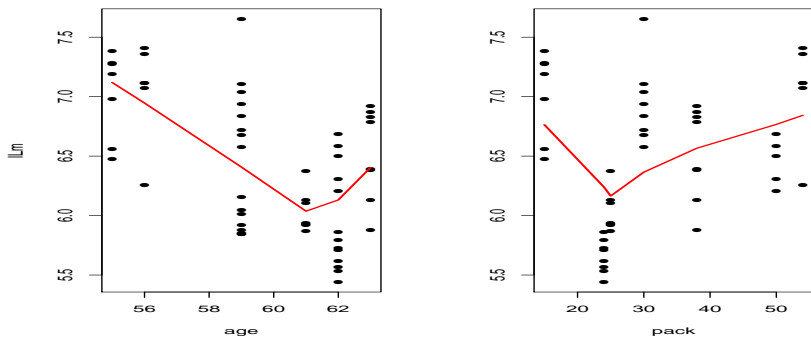


Figure 2: lowess fitting for $\log(\text{lLm})$ with packyears and age respectively

Random Intercepts of $\log(L_m)$

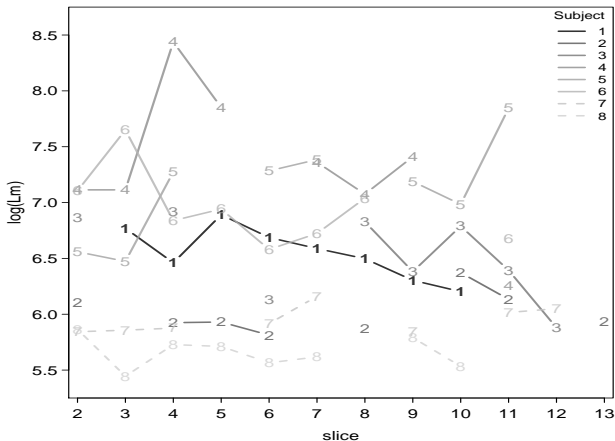


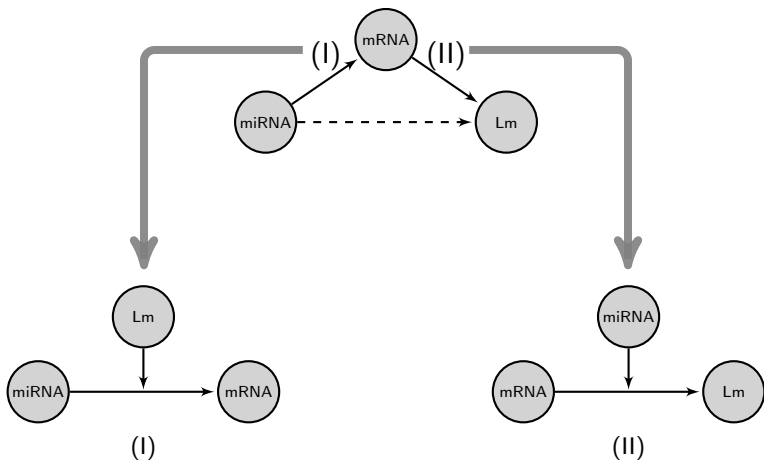
Figure 3: Spaghetti plot for $\log(L_m)$ with slice for each subject. Slice 2 is the apex in lung and slice 13 is the bottom in lung. Subject 1 to 6 represent patients with COPD and subject 7 and 8 represent donors without COPD.

Model Selection

Parsimonious Model for $\log(L_m)$	logLik	AIC
COPD + packyears* + age* + sex + slice	-0.70	41.39
COPD + packyears* + age* + sex	-8.14	34.28
COPD + packyears* + age*	-13.86	43.72
COPD + packyears*	-17.71	47.41
COPD	-18.92	45.84

Table 1: Fits of parsimonious longitudinal models for $\log(L_m)$ with various demographic variable including packyears(number of packs of cigarettes consumed each year), age, sex and slice. Note $\text{age}^* = \text{age} + (\text{age} - 61)_+$ and $\text{packyears}^* = \text{packyears} + (\text{packyears} - 25)_+$ are used in model fittings, motivated by the exploratory study in figure 2.

Weight-adjusted p-value method



Weight-adjusted p-value method

Step 1. Obtain p-values for 397 miRNAs, denoted by $p_j, j = 1, 2, \dots, 397$, using likelihood ratio test based on the linear mixed effect models (1), (2).

$$lLm = u + \mathbf{x}'\boldsymbol{\beta} + \varepsilon, \quad (1)$$

$$lLm = u + \mathbf{x}'\boldsymbol{\beta} + \beta_{\text{miR}_j, Lm} \text{miRNA}_j + \varepsilon, \quad (2)$$

Weight-adjusted p-value method

Step 2. Calculate the weight matrix \mathbf{W} in which each element w_{jk} represents to the extent that the j th miRNA is associated with the emphysema severity ILm by regulating the k th mRNA.

$$w_{jk} = \underbrace{\left(\frac{\hat{\beta}_{miR_j, mR_k}}{SE(\hat{\beta}_{miR_j, mR_k})} \right)^2}_{w_{miR_j, mR_k}} \times \underbrace{\left(\frac{\hat{\beta}_{mR_k, ILm}}{SE(\hat{\beta}_{mR_k, ILm})} \right)^2}_{w_{mR_k, ILm}}$$

Where $\hat{\beta}_{miR_j, mR_k}$ and $\hat{\beta}_{mR_k, ILm}$ are obtained from the linear mixed effect model (3) and (4), accordingly.

$$mRNA_k = u + \mathbf{x}'\boldsymbol{\beta} + \beta_{miR_j, mR_k} miRNA_j + \beta_{ILm} ILm + \varepsilon \quad (3)$$

$$ILm = u + \mathbf{x}'\boldsymbol{\beta} + \beta_{mR_k, ILm} mRNA_k + \beta_{miR_j} miRNA_k + \varepsilon \quad (4)$$

Weight-adjusted p-value method

$$\mathbf{W} = \left[\begin{array}{ccccc}
 \overbrace{w_{1,1} & w_{1,2} & w_{1,3} & \dots & w_{1,22011}}^{\text{mRNA}} \\
 w_{2,1} & w_{2,2} & w_{2,3} & \dots & w_{2,22011} \\
 \vdots & \vdots & \vdots & \ddots & \vdots \\
 w_{396,1} & w_{396,2} & w_{396,3} & \dots & w_{396,22011}
 \end{array} \right] \left. \vphantom{\begin{array}{c} \\ \\ \\ \\ \end{array}} \right\} \text{microRNA}$$

step 3. Calculate $w_{\text{miR}_j} = \max_k w_{jk}$ and assign the average scaled weight $w_{\text{miR}_j}^* = \frac{w_{\text{miR}_j}}{\overline{w_{\text{miR}}}}$ to the j th miRNA, such that $w_{\text{miR}_j}^* > 0$ and the average of all weights $\overline{w^*}$ is 1 and $\overline{w_{\text{miR}}}$ is the average of all w_{miR_j} . Then, the weight-adjusted p-value for the j th miRNA is

$$\text{P}_{\text{adj}} = \frac{p_j}{w_{\text{miR}_j}^*}$$

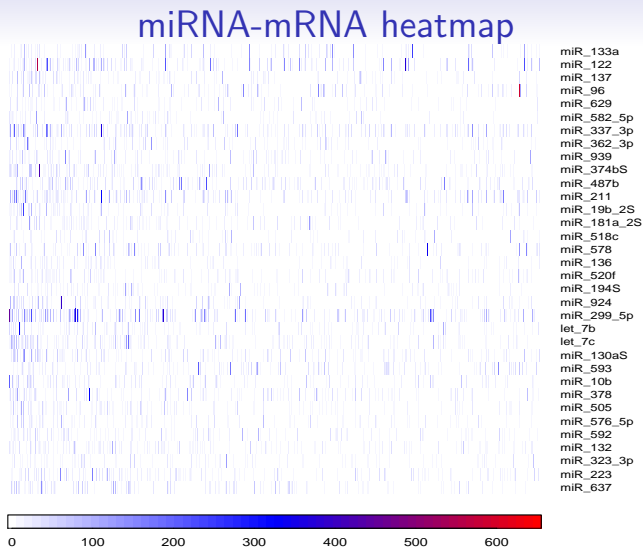
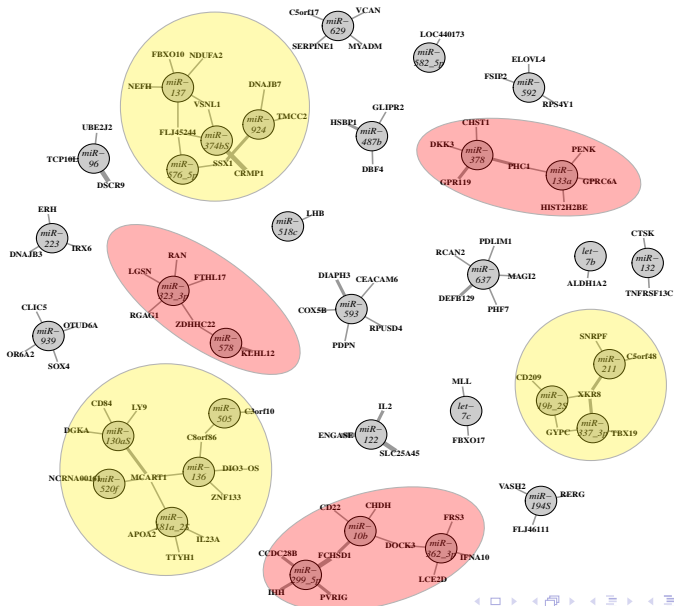


Figure 4: Weight matrix. Rows are 34 top significant miRNAs in terms of p-values. Columns are 758 mRNAs which are the combined top 5 mRNAs that are most associated with the 397 miRNAs.

miRNA-mRNA regulation network



Gene Enriched Analysis

	ID	PathName	Pvalue	Odds	Expected
1	05320	Autoimmune thyroid disease	0.023	9.346	0.238
2	04672	Intestinal immune network for IgA production	0.024	9.117	0.243
3	05012	Parkinson's disease	0.025	5.265	0.633
4	04060	Cytokine-cytokine receptor interaction	0.041	3.374	1.324
5	04630	Jak-STAT signaling pathway	0.043	4.198	0.786

Table 2: Enriched KEGG pathway using 145 top genes regulated by the 34 microRNAs that are most associated with ILM.

Conclusion

1. Proposed a new linear mixed model to analyze COPD data by integrating the genotype and phenotype information.
2. Used the weight-adjusted p-value method to detect the significant microRNAs that are most associated with COPD while controlling the nominal Family-Wise Error Rate.
3. Discovered a new sparse regulation network between microRNA and mRNA that might be beneficial to unveil the pathogenesis of COPD.
4. Provided a new way to discover the regulation network between microRNA and mRNA for cancer disease.

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The End

Thank you!