

Unveiling the Impact of Social and Environmental Determinants of Health on Lung Function Decline in Cystic Fibrosis through Data Integration using the US Registry

International Society for Clinical Biostatistics

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Introduction: Motivation

A lot of information is available
→ Electronic medical records

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 - Electronic medical records

Cystic Fibrosis

- genetic disorder affecting the lungs, pancreas, and other organs
- > 75 percent of people with CF are diagnosed by age 2

US Cystic Fibrosis Registry

- ◊ >23,000 patients
- ◊ >1,400,000 observations (on average >10 years of follow-up)



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Introduction: Motivation

A lot of information is available

- Electronic medical records

Different types of information

- Baseline characteristics: Sex, F508del, SESlow, Enzymes
- Biomarkers: FEV₁ % pred
- Nutritional status: BMI percentile
- Social and environmental determinants: Deprivation index

Introduction: Motivation

Deprivation index

- Socioeconomic variables from the American Community Survey (ACS): capture “community deprivation”
 - ◊ Principal components analysis of six different 2015 ACS measures
 - ◊ “Deprivation Index”: the first component explains over 60% of the total variance
 - ◊ Rescaling and normalizing forces the index to range from 0 to 1, with a higher index being more deprived



Cole Brokamp, Andrew F. Beck, Neera K. Goyal, Patrick Ryan, James M. Greenberg, Eric S. Hall. Material Community Deprivation and Hospital Utilization During the First Year of Life: An Urban Population-Based Cohort Study. *Annals of Epidemiology*. 30. 37-43. 2019

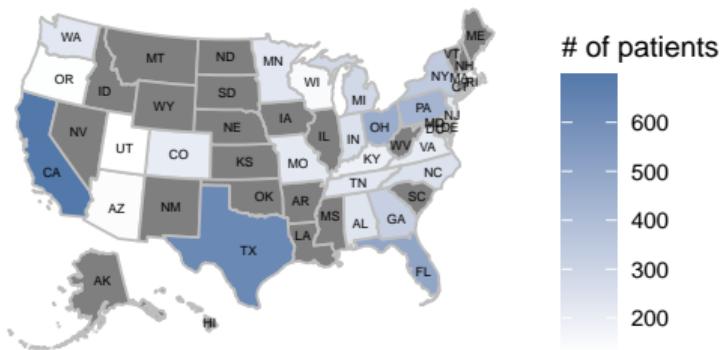
https://geomarker.io/dep_index/

Introduction: Research question

Can we integrate registry data with social and environmental determinants of health to improve the accuracy of disease progression prognostication?

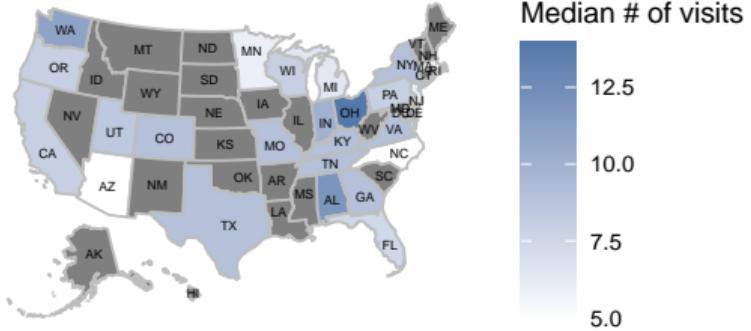
Introduction: Descriptive statistics

Baseline information



of patients

Value
200
300
400
500
600

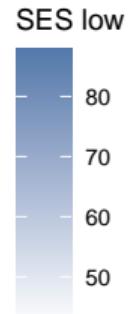
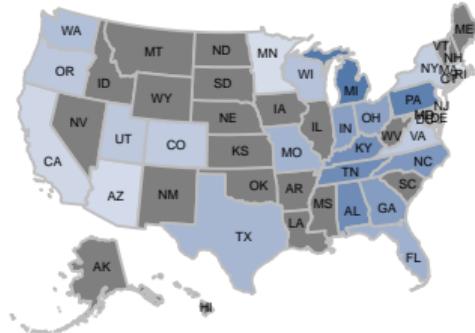
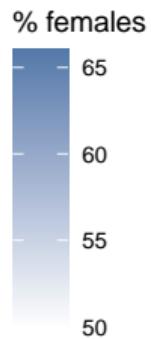
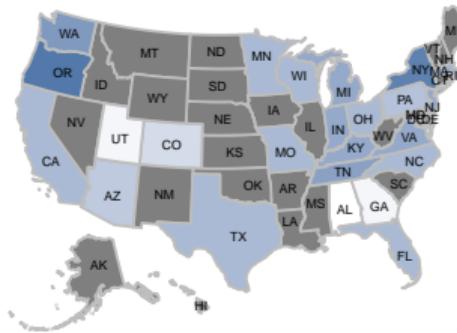


Median # of visits

Value
5.0
7.5
10.0
12.5

Introduction: Descriptive statistics

Baseline information

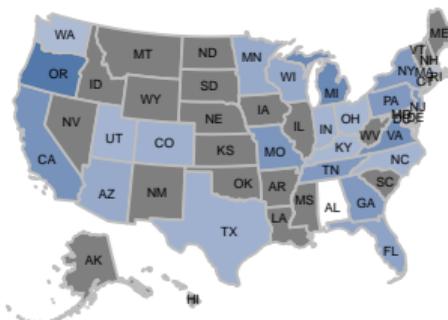
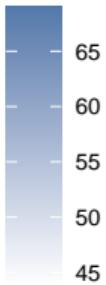


Introduction: Descriptive statistics

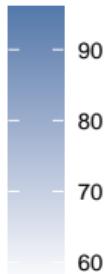
Baseline information



F508del Homozygous



Enzymes Yes



Introduction: Descriptive statistics

Lung function

Introduction: Descriptive statistics

Social and environmental determinants of health

Methods: Multivariate Mixed Models

DepIndex: $y_{2i}(t) = x_{2i}^\top(t)\beta_1 + z_{2i}(t)^\top b_{2i} + \epsilon_{2i}(t)$

FEV1%pred: $y_{1i}(t) = x_{1i}^\top(t)\beta_1 + z_{1i}(t)^\top b_{1i} + \epsilon_{1i}(t)$

where

$$\diamond \quad b_i^\top = (b_{1i}^\top, b_{2i}^\top) \sim N(0, D)$$

Cannot directly measure the strength of the association and lack clinical relevance

Methods: Multivariate Mixed Models

$$\text{DepIndex: } y_{2i}(t) = m_{2i}(t) + \epsilon_{2i}(t) = x_{2i}^\top(t)\beta_1 + z_{2i}(t)^\top b_{2i} + \epsilon_{2i}(t)$$

$$\text{FEV1\%pred: } y_{1i}(t) = x_{1i}^\top(t)\beta_1 + z_{1i}(t)^\top b_{1i} + \alpha_{S2} \int_0^t m_{2i}(s)ds + \epsilon_{1i}(t)$$

where

$$\diamond \quad b_i^\top = (b_{1i}^\top, b_{2i}^\top) \sim N(0, D)$$

R package: <https://github.com/ERandrinopoulou/multiLME>

Methods: Multivariate Mixed Models

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FEV1%pred: $y_{1i}(t) = x_{1i}^\top(t)\beta_1 + z_{1i}(t)^\top b_{1i} + \alpha_{S2} \int_{t-d}^t m_{2i}(s)ds + \epsilon_{1i}(t)$

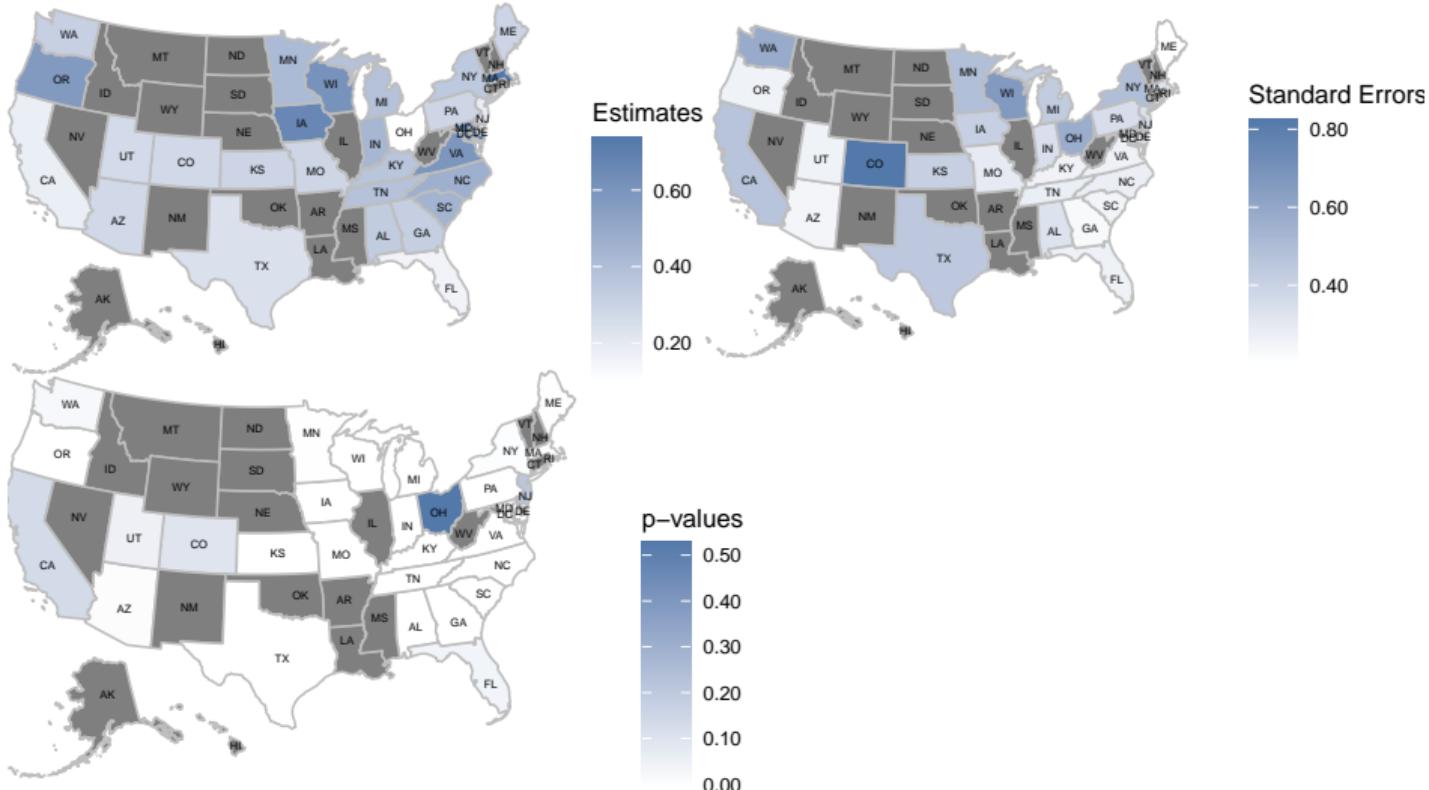
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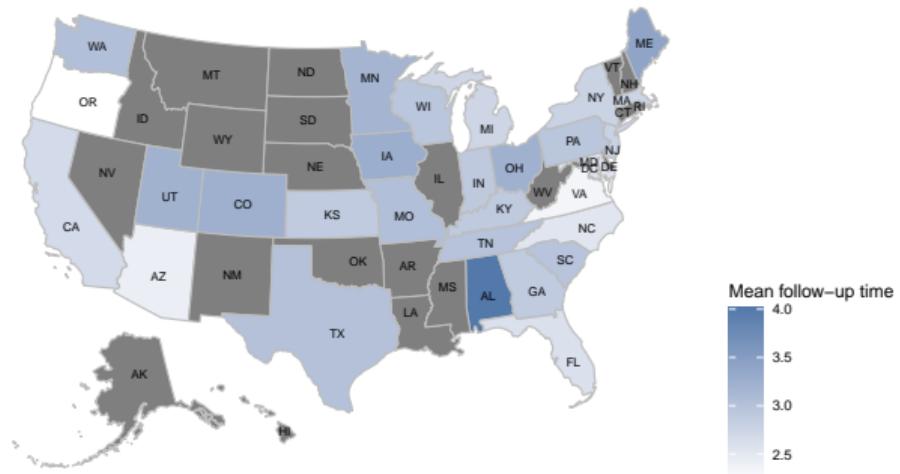
Results: Multivariate Mixed Models

Estimate: for 0.1 unit increase in the area under the curve of the deprivation index



Results: Multivariate Mixed Models

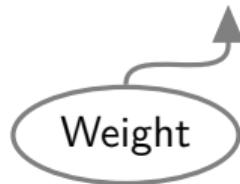
Descriptive statistics: Baseline information



Methods: Multivariate Mixed Models

DepIndex: $y_{2i}(t) = m_{2i}(t) + \epsilon_{2i}(t) = x_{2i}^\top(t)\beta_1 + z_{2i}(t)^\top b_{2i} + \epsilon_{2i}(t)$

FEV1%pred: $y_{1i}(t) = x_{1i}^\top(t)\beta_1 + z_{1i}(t)^\top b_{1i} + \alpha_{S2} \frac{1}{t} \int_0^t m_{2i}(s)ds + \epsilon_{1i}(t)$

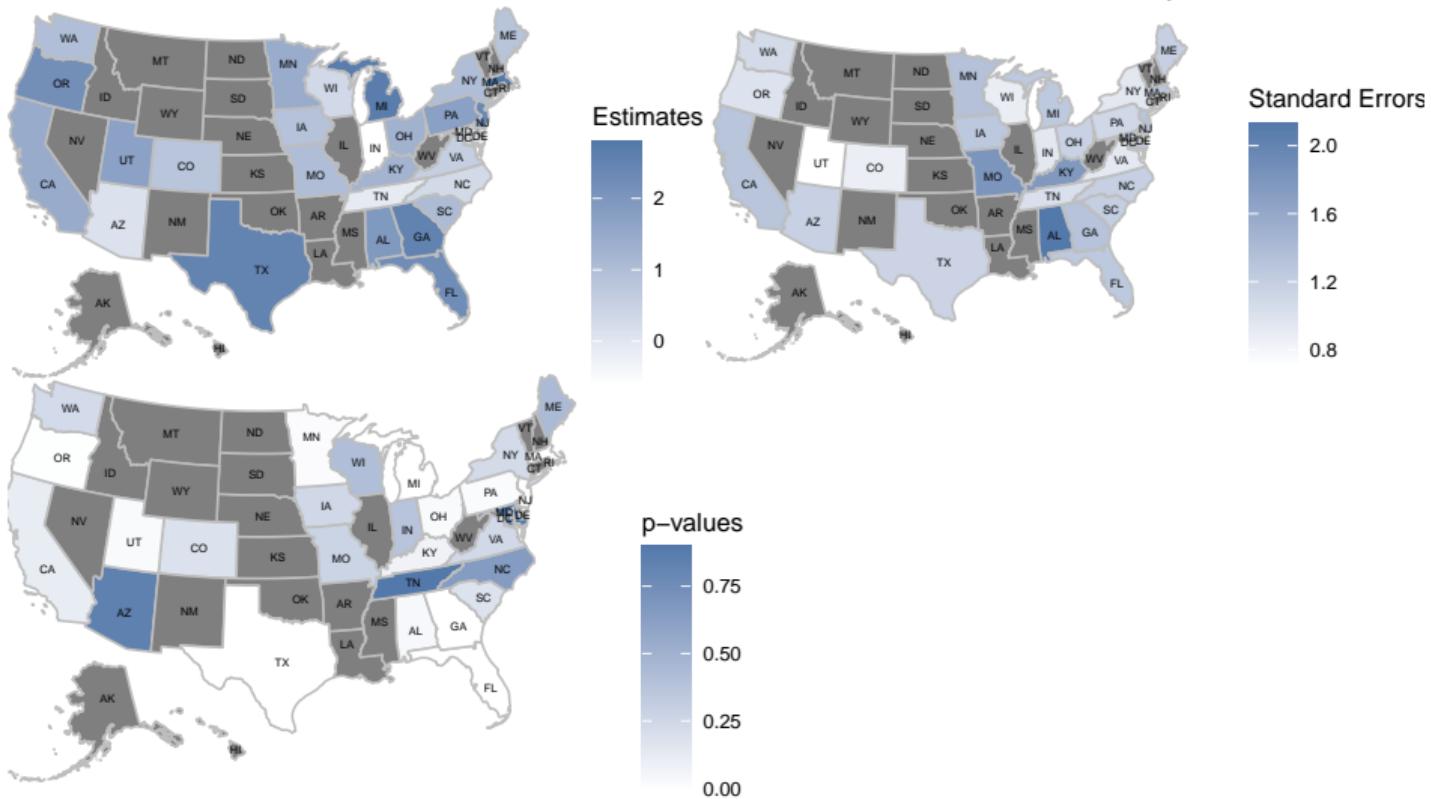


where

◇ $b_i^\top = (b_{1i}^\top, b_{2i}^\top) \sim N(0, D)$

Results: Multivariate Mixed Models

Estimate: for 0.1 unit increase in the normalized area under the curve of deprivation index



Results: Multivariate Mixed Models

- When the exposure time is short (e.g., 2 or 5 years), the association becomes weaker with some differences between the states.

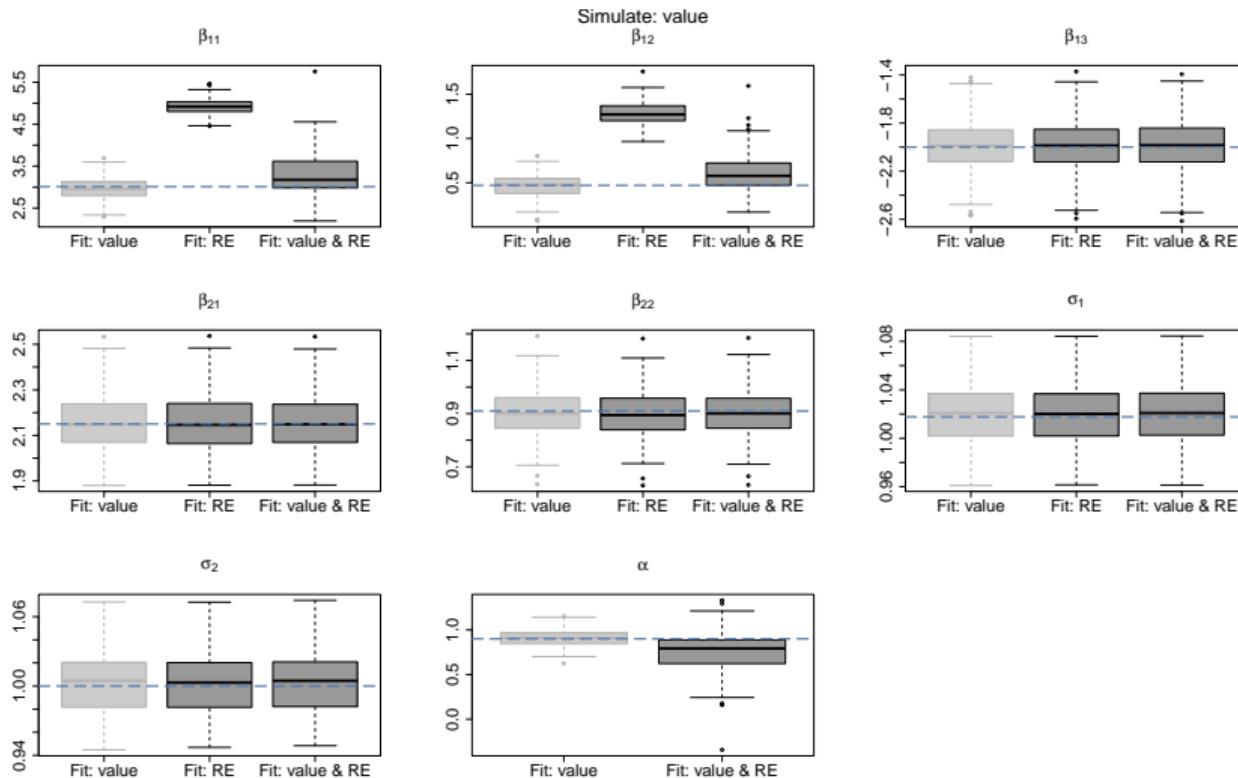
Sensitivity analysis:

- When 2 and 5 years are assumed, there is a stronger association when the diagnosis age is *below 18*. When 10 and 15 years are assumed, there is a stronger association when the diagnosis age is *above 12*.

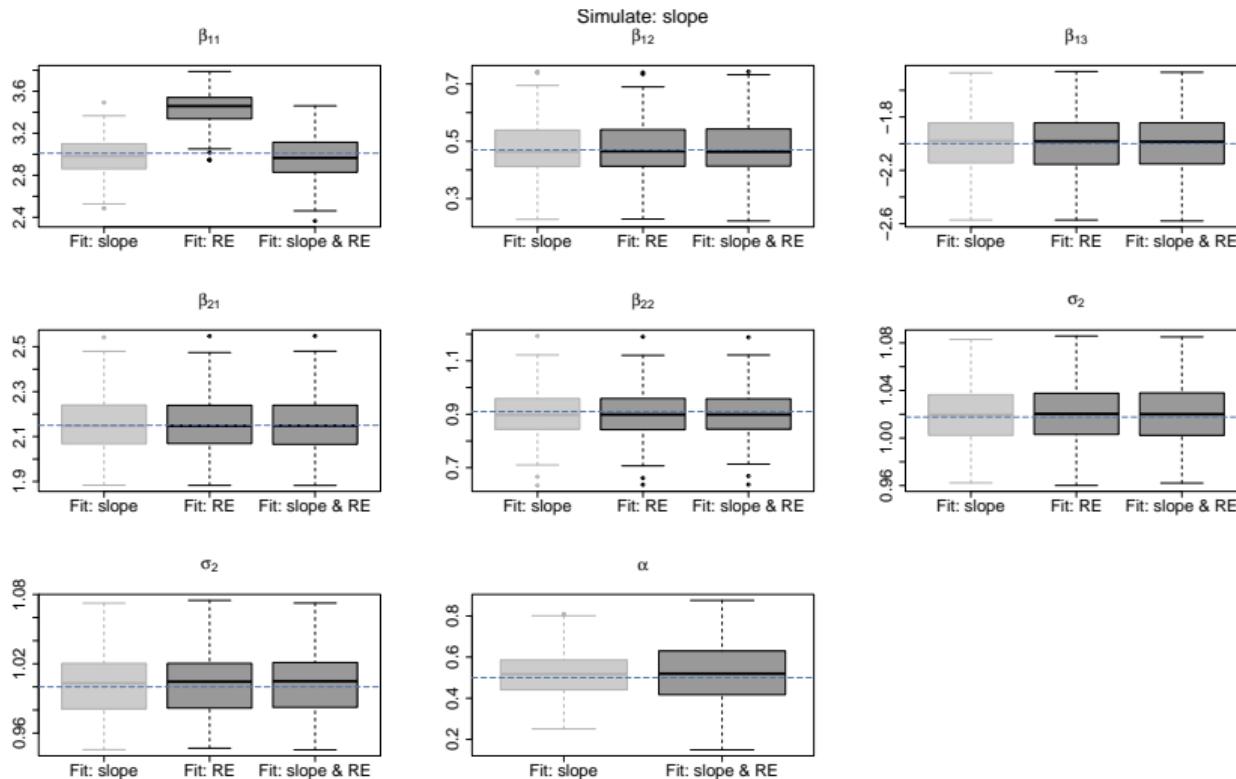
Simulation: Set up

- **Real-world data**: defining the connection between the different longitudinal outcomes is a difficult task
- Bias: when we ignore or over specify the relationship between different data sources.
 - **Simulate**: different forms of association (value/area/slope)
 - ◊ **Fit**: value/area/slope
 - ◊ RE
 - ◊ value/area/slope + RE

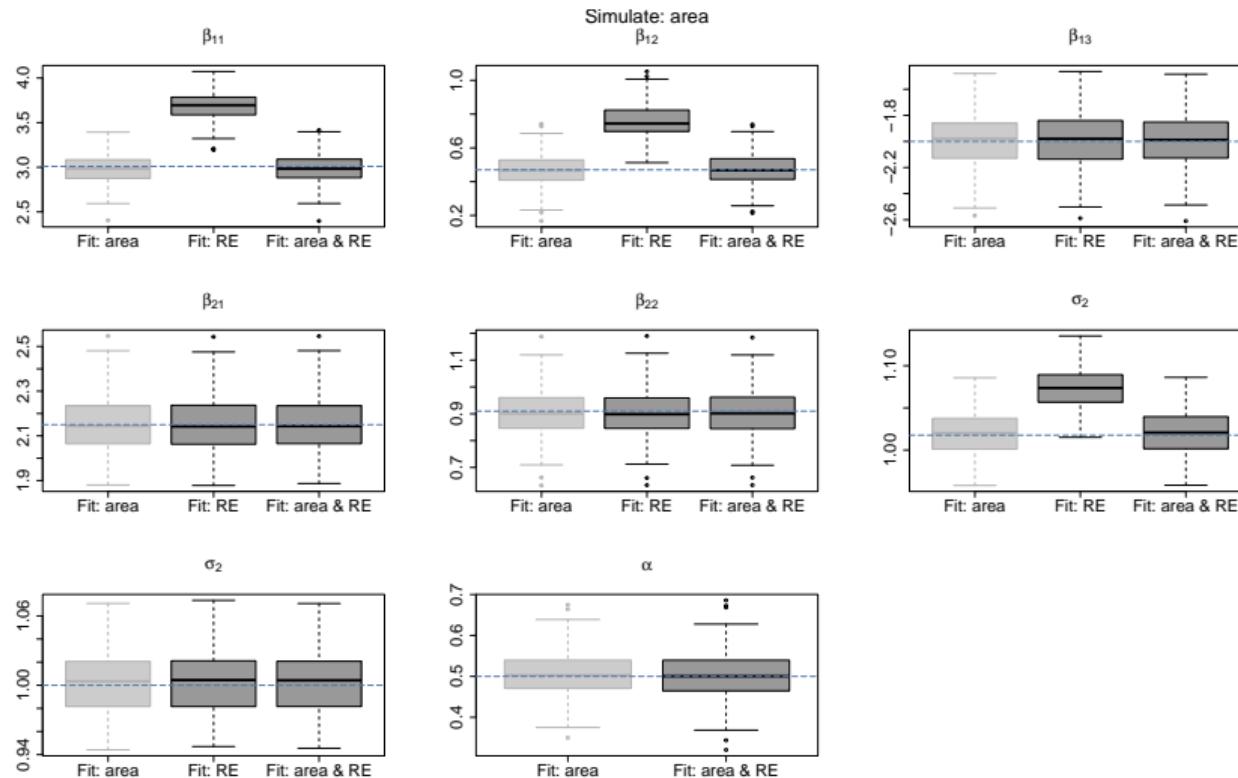
Simulation: Results



Simulation: Results



Simulation: Results



Conclusion

- A lot of data is available
- Better treatment and monitoring strategies if all information is used
- Challenge in combining different types of information
- Investigate other social and environmental determinants of health

Thank you for your attention!