# A Very Basic Intro to SLCMA



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### 99 PROBLEMS

#### WHAT WE HAVE

- Several large panel studies
- Many years of data
- A plethora of data types
- Really interesting questions

Year 3

Yearl

Birth

Year 5

#### WHAT WE NEED

- Way to use our vast data to test hypotheses across the life course
  - Systematic to avoid false-positive results
  - Efficient to accommodate analysis of highdimensional data
  - Easy to use



Year 9





### WHAT IS SLCMA?

STRUCTURED

LIFE

COURSE

MODELLING

**Approach** 



#### WHICH LIFE COURSE HYPOTHESIS BEST FITS OUR DATA?



### SLCMA R PACKAGE



#### • R Package (slcma)

- Created by Andrew Smith
- Under review by CRAN
- I'll be talking in the broadest terms
  - No code
  - Reach out for more info

### SLCMA STEPS

- 1. Fit a regression model for each single life course hypothesis of interest, as well as groups of hypotheses
  - Manually: Fit a model for each hypothesis, as well as compound hypotheses
  - **slcma:** Uses LARS to fit each model

#### 2. Measure the goodness-of-fit of each model and select the best one

- Manually: Create an elbow plot
- **slcma:** Generates an elbow plot with a single command
- 3. Calculate appropriate *p*-values for the selected model
  - Manually: Apply a Bonferroni correction
  - **slcma:** Uses fixed LASSO inference or max-|t| test to correct *p*-values



### HIGH THROUGHPUT

#### **METHYL ARRAY DATA**

- A LOT OF MODELS!!
  - $850K CpGs \times 5 Hyp = 4.25M Regs$
- SLCMA can select the best-fitting hypothesis for each probe
  - Compound hypotheses are difficult
- SLCMA can typically find more associations than cross-sectional or ever-exposed EWAS models
  - Multiple-testing correction raises significance threshold significantly



# SLCMA HYPOTHESES

BIG 3
Sensitive Periods
Accumulation
Recency

- OTHERS
- Mobility
- Change
- Always Exposed
- Ever Exposed



# HYPOTHESES (BIG 3)

#### **SENSITIVE PERIODS (**SP at $t_i$ **)**

- The developmental timing of an exposure has the strongest effect on the outcome at a specific time point due to heightened levels of plasticity or reprogramming
- Just the exposure variable at each age
  - $SP_j = x_j$
- Can be continuous or binary



# HYPOTHESES (BIG 3)

#### **ACCUMULATION (***Acc***)**

- Every additional time point of exposure affects the outcome in a dose-response manner, independent of the exposure timing
- Add up the exposure variable across ages
  - Acc =  $\sum_{j=1}^{m} x_j$
- Can be continuous or binary



# HYPOTHESES (BIG 3)

#### **RECENCY (Rec)**

- More proximal exposures (closer in time to the of the outcome) are more strongly linked to the outcome than are more distal exposures
- Add up the products of each exposure variable multiplied by its age of observation

• Rec =  $\sum_{j=1}^{m} (x_j t_j)$ 

• Can be continuous or binary













Test each hypothesis independently. Accumulation is the best fit.



#### $Y \sim Acc + SP_1$

MODELS Y ~ Acc

Y ~ Acc + SP<sub>2</sub>

 $Y \sim Acc + SP_3$ 

#### Y ~ Acc + Rec

Test Accumulation w/ each remaining hypothesis. Accumulation + Sensitive Period 1 is the best fit.



 $\frac{Y \sim Acc + SP_1 + SP_2}{Y \sim Acc + SP_1 + SP_3}$ 

 $Y \sim Acc + SP_1 + Rec$ 

Test Accumulation + Sensitive Period 1 w/ each remaining hypothesis. Accumulation + Sensitive Period 1 + Sensitive Period 3 is the best fit.

MODELS

Y ~ Acc

 $Y \sim Acc + SP_1$ 

**Y** Acc + SP<sub>1</sub> + SP<sub>3</sub> + Rec

 $+ Acc + SP_1 + SP_3 + SP_7$ 

MODELS Y ~ Acc Y ~ Acc +  $SP_1$ Y ~ Acc +  $SP_1$  +  $SP_3$ 

Model fit is identical in both remaining models. Accumulation & Recency are linear combinations of SP<sub>1</sub>, SP<sub>2</sub>, and SP<sub>3</sub>.

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# Fitting Models

MODELS Y ~ Acc  $Y \sim Acc + SP_1$  $Y \sim Acc + SP_1 + SP_3$ **HOW DO WE CHOOSE?** 





### MULTIPLE TESTING

#### **ONE HYPOTHESIS IN MODEL, BUT WE TESTED FIVE**

- We need to correct for multiple tests
- Naive Calculation
  - Inflated FWER; Biased *p*-values
  - Fast computation time
- Bonferroni Correction
  - FWER Controlled; Unbiased *p*-values
  - Overly conservative

• Fixed LASSO Inference

- FWER Controlled; Unbiased *p*-values
- Slow computation time
- Max-|t|-Test
  - FWER Controlled; Unbiased *p*-values
  - Slow computation time

### **MOBILITY** $(t_j \text{ to } t_k)$

- The direction of change in exposure status between 2 time periods, rather than the absolute state at each individual time point, affects the outcome
- Upward mobility equal to 1 if exposed after unexposed, 0 if not
  - $Mob_{jk}^+ = (1-x_j)x_k$
- Downward mobility equal to 1 if unexposed after exposed, 0 if not

• 
$$Mob_{jk}^- = x_j(1-x_k)$$



### CHANGE $(t_j \text{ to } t_k)$

- The change in exposure status between 2 time periods, rather than the absolute state at each individual time point, affects the outcome
- Subtract the exposure at the later age from the earlier one

• 
$$Chn_{jk} = x_k - x_j$$



#### **ALWAYS EXPOSED**

- Constant exposure over time affects the outcome
- Equal to 1 if always exposed, 0 if not
  - $Alw = \prod_{j=1}^{m} (x_j)$



#### **EVER EXPOSED**

- Exposure at any time point affects the outcome
- Equal to 1 if ever exposed, 0 if not
  - $Evr = 1 \prod_{j=1}^{m} (1 x_j)$



### THEORY IS KEY

- SLCMA is inseparable from theory
- Better tools for machine learning outside of the theoretical framework
- Choose theoretically motivated hypotheses
  - Each hypothesis tested inflates p-values
  - Be parsimonious
- You can construct your own hypotheses



### THANK YOU FEEL FREE TO REACH OUT WITH ANY QUESTIONS JOSHUA A. GOODE jagoode@umich.edu

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