

A methodical review of the causal role of socioeconomic determinants of health disparities

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Intuitive model of Health Disparities

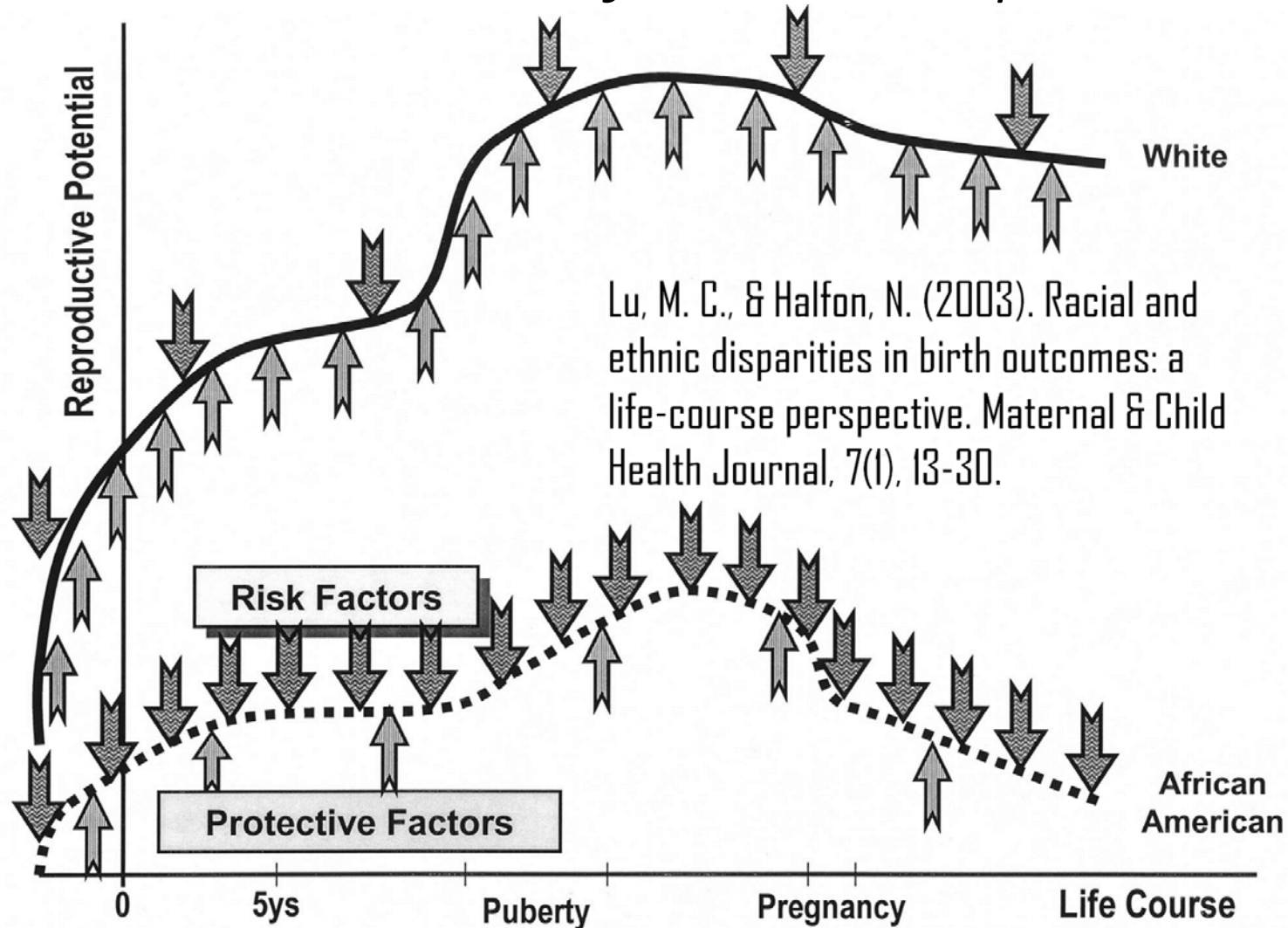


Fig. 1. How differential exposures to risk factors (downward arrows) and protective factors (upward arrows) over the life course affect developmental trajectories and contribute to disparities in birth outcomes. The lower reproductive potential of African American women, relative to White women, results from their cumulative exposure to more risk factors and less protective factors across the life span, particularly during sensitive periods of development.

Statistics -> Modeling -> Causal modeling

	Statistical Concepts	Causal Concepts
1	Correlation	Randomization
2	Regression	Influence
3	Dependence	Effect
4	Conditional independence	Confounding
5	Likelihood	“Holding constant”
6	Collapsibility	Disturbance
7	Propensity score	Error terms
8	Risk ratio	Structural coefficients
9	Odds ratio	Spurious correlation
10	Marginalization	Faithfulness/stability
11	Conditionalization	Instrumental variables
12	“Controlling for”	Intervention
13		Explanation
14		Attribution

Statistics vs. Modeling

A baby step: separate out in analytical work:

1. Modeling
2. Statistical testing

I.

A t-test can be seen as a

1. Model (2-group 1-continuous variable)
2. Test (Student's t: ideal distribution vs. real/sample)

II.

A correlation can be seen as a

1. Model (1-group 2-continuous variable)
2. Test (z or t test of ρ significance)

III.

A regression can be seen as a

1. Model (directional: 1-group 2-continuous variable)
2. Test (of β or of R^2)

Modeling vs. statistics

#R made nicely clear one can build a data-free model

```
install.packages("MIIVsem")
```

```
model.miiv1 <- ' DV~ IV1 + IV2'
```

#Model and Data are 2 ingredients: fitting is a 3rd

```
fit.miiv.real <- miive(model.miiv1, data.real)
```

#One can simulate data off a model too

```
data.sim.1 <- simulateData(model = model.miiv1,  
sample.nobs = 1000, seed = 123)
```

```
fit.miiv.sim <- miive(model.miiv1, data.sim.1)
```

#dagitty can plot the model too

SES and its place in the causal scheme

- Socioeconomic position (SEP) [1]

* “The research goals of measuring SEP: monitoring aetiology and confounding.

- The most obvious purpose in measuring SEP is to describe and monitor the social distribution of a disease in order to inform health policy.

- The second purpose for measuring SEP relates to explaining the *causal mechanisms* through which SEP *generates* health differences.

- The third purpose of measuring SEP in health-related research is to statistically adjust for socioeconomic circumstances when another exposure is the main focus of interest.

In this context, it is crucial to fully account for *confounding* effects due to socioeconomic conditions.” [2]:4-5.

1. Lu, M. C., & Halfon, N. (2003). Racial and ethnic disparities in birth outcomes: a life-course perspective. *Maternal & Child Health Journal*, 7(1), 13-30.

2. Galobardes, B., Lynch, J., & Smith, G. D. (2007). Measuring socioeconomic position in health research. *British medical bulletin*, 81(1), 21.

SES issues

Terminology

- Socioeconomic position (SEP) [1]

- How independent are SES indicators?
 - Wealth \neq income, e.g.
 - “even at similar levels of income, non-white minorities lag behind whites in total assets and net worth (the difference between assets and liabilities)” [1]

- Should one combine them into some weighted total?
- Is there a latent SES?
- Objective vs. subjective SES.

1. Pollack, C. E., Chideya, S., Cubbin, C., Williams, B., Dekker, M., & Braveman, P. (2007). Should health studies measure wealth?: A systematic review. *American Journal of Preventive Medicine, 33*(3), 250-264.

SES & structures

“The validity of common adjustment strategies when estimating the outcome distribution under hypothetical interventions of the exposure is potentially compromised by **structured relations** between covariates, observed and unobserved.”

“conditioning on covariates without regard to **structured interrelations** among the relevant quantities (measured and unmeasured) can lead to biased estimation of casual effects”[1:157-8]

What is A 'structure'

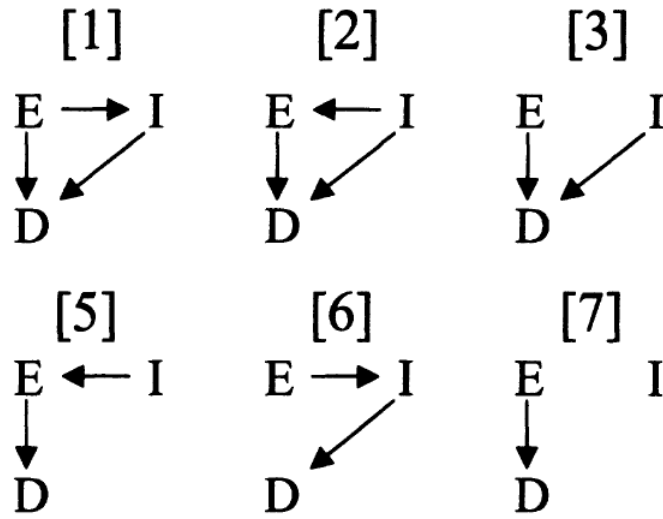
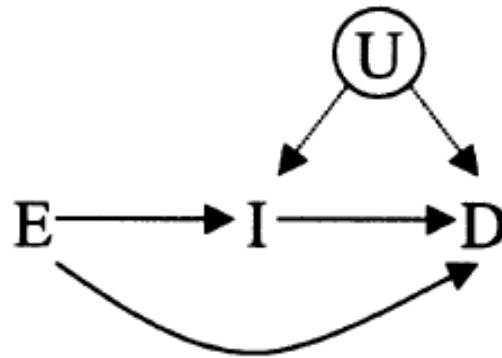


FIGURE 2.

Scenario #1
(Graph [1G])



Scenario #2
(Graph [6E])

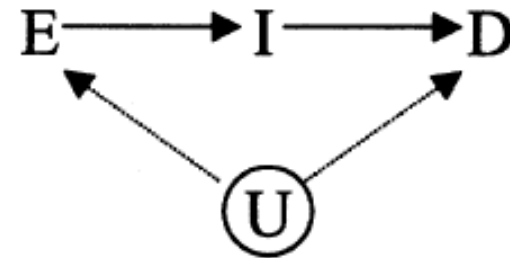


FIGURE 4.

E = Education; I = Income; D = Death (Mortality)

Kaufman – structure analyzed (1 age group)

Appendix 2

TABLE A2.1

Income	Education			
	Low	Mid	High	Total
Low	ALIVE: 2350	ALIVE: 918	ALIVE: 315	ALIVE: 3583
	DEAD: 1024	DEAD: 322	DEAD: 116	DEAD: 1462
Mid	ALIVE: 2931	ALIVE: 2366	ALIVE: 1038	ALIVE: 6335
	DEAD: 790	DEAD: 524	DEAD: 260	DEAD: 1574
High	ALIVE: 2379	ALIVE: 3666	ALIVE: 4417	ALIVE: 10462
	DEAD: 445	DEAD: 634	DEAD: 594	DEAD: 1673
Total	ALIVE: 7660	ALIVE: 6950	ALIVE: 5770	ALIVE: 20380
	DEAD: 2259	DEAD: 1480	DEAD: 970	DEAD: 4709

E = Education; I = Income; D = Death (Mortality)

Kaufman – structure analyzed (1 age group)

Step 5: Standard Adjustment for Effect of E given I:

$$\sum_i (P(d|e, i) P(i)):$$

$$P(\text{dead}|\text{SET}[E = L]) = (1024/(2350 + 1024)) \cdot (0.201) + \\ (790/(2931 + 790)) \cdot (0.315) + \\ (445/(2379 + 445)) \cdot (0.484) = 0.204$$

Step 7: Structured Adjustment for Scenario 2:

$$\sum_i [P(i|e) \sum_{e'} (P(d|e', i) P(e'))]:$$

	$I = L$	$I = M$	$I = H$				
	<u>Step 3</u>	<u>Step 6</u>	<u>Step 3</u>	<u>Step 6</u>	<u>Step 3</u>	<u>Step 6</u>	
$P(\text{dead} \text{SET}[E = L])$	(0.340)	$\cdot (0.280)$	$+$	(0.375)	$\cdot (0.199)$	$+$	$(0.285) \cdot (0.144) = 0.211$
$P(\text{dead} \text{SET}[E = M])$	(0.147)	$\cdot (0.280)$	$+$	(0.343)	$\cdot (0.199)$	$+$	$(0.510) \cdot (0.144) = 0.183$
$P(\text{dead} \text{SET}[E = H])$	(0.063)	$\cdot (0.280)$	$+$	(0.193)	$\cdot (0.199)$	$+$	$(0.743) \cdot (0.144) = 0.163$

SES as a latent factor

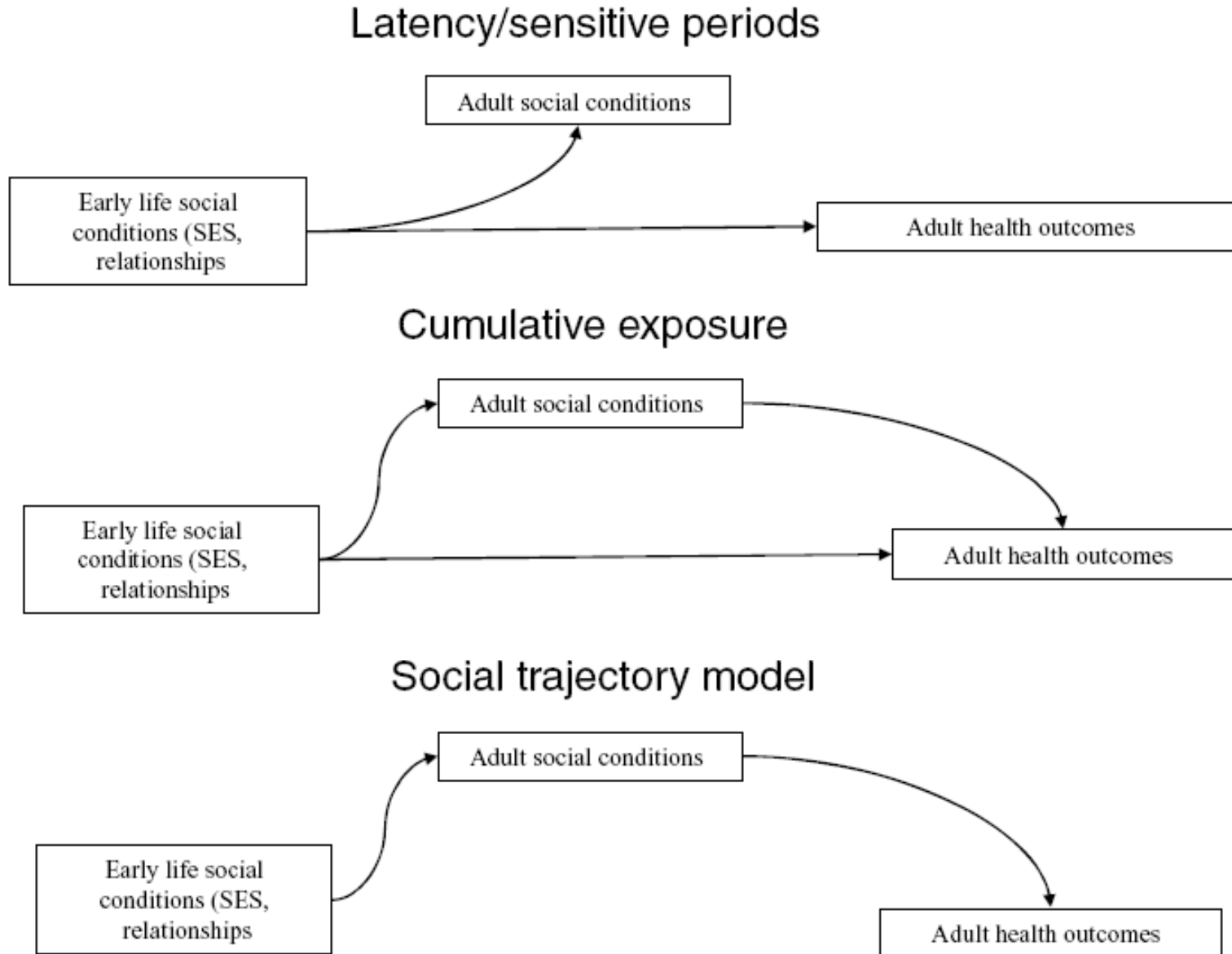
Table 1 38 young adult (Add Health Wave III; 2000–2001) SES indicator variables used to generate SES factors, listed by three domains of SES (Material, Human, and Social Capital).

<p><u>MATERIAL CAPITAL</u></p> <p><u>Income sources (no/yes)</u> Wages, including tips/bonus Interest from stocks, bonds Income from family/friends</p> <p><u>Personal economics</u> Personal income in 2001 Own residence (no/yes) Own vehicle (no/yes)</p> <p><u>Information and financial access (no/yes)</u> Own/access to computer Have email account Have checking account Have credit card Have savings account Have shares of stock Have student loan Have credit card debt</p> <p><u>Economic hardship in the last year (no/yes)</u> Without telephone service Unable to pay rent/mortgage Gas/electricity/oil turned off Unable to afford doctor Evicted for not paying rent</p>	<p><u>Public assistance (no/yes)</u> Housing assistance Food stamps AFDC, welfare Ever received assistance other than food stamps</p> <p><u>Miscellaneous</u> Number of months of health insurance in past year Currently living with parents (no/yes)</p> <p><u>HUMAN CAPITAL</u></p> <p><u>Education</u> Highest grade attained HS diploma (no/yes) BA degree (no/yes) In school part/full time (no/yes) In 4-year college (no/yes)</p> <p><u>Labor experience</u> Number of jobs Job description</p> <p><u>SOCIAL CAPITAL</u> Community activities Volunteer experience Organ donor (no/yes) Registered to vote (no/yes) Voted in 2000 (no/yes) Political affiliation (no/yes)</p>
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Scharoun-Lee, M., Adair, L. S., Kaufman, J. S., & Gordon-Larsen, P. (2009). Obesity, race/ethnicity and the multiple dimensions of socioeconomic status during the transition to adulthood: A factor analysis approach. *Social Science & Medicine*, 68(4), 708-716. doi:<https://doi.org/10.1016/j.socscimed.2008.12.009>

Figure 3

Three life course models of disease: latency, cumulative exposures, and social trajectories.



Subjective SES

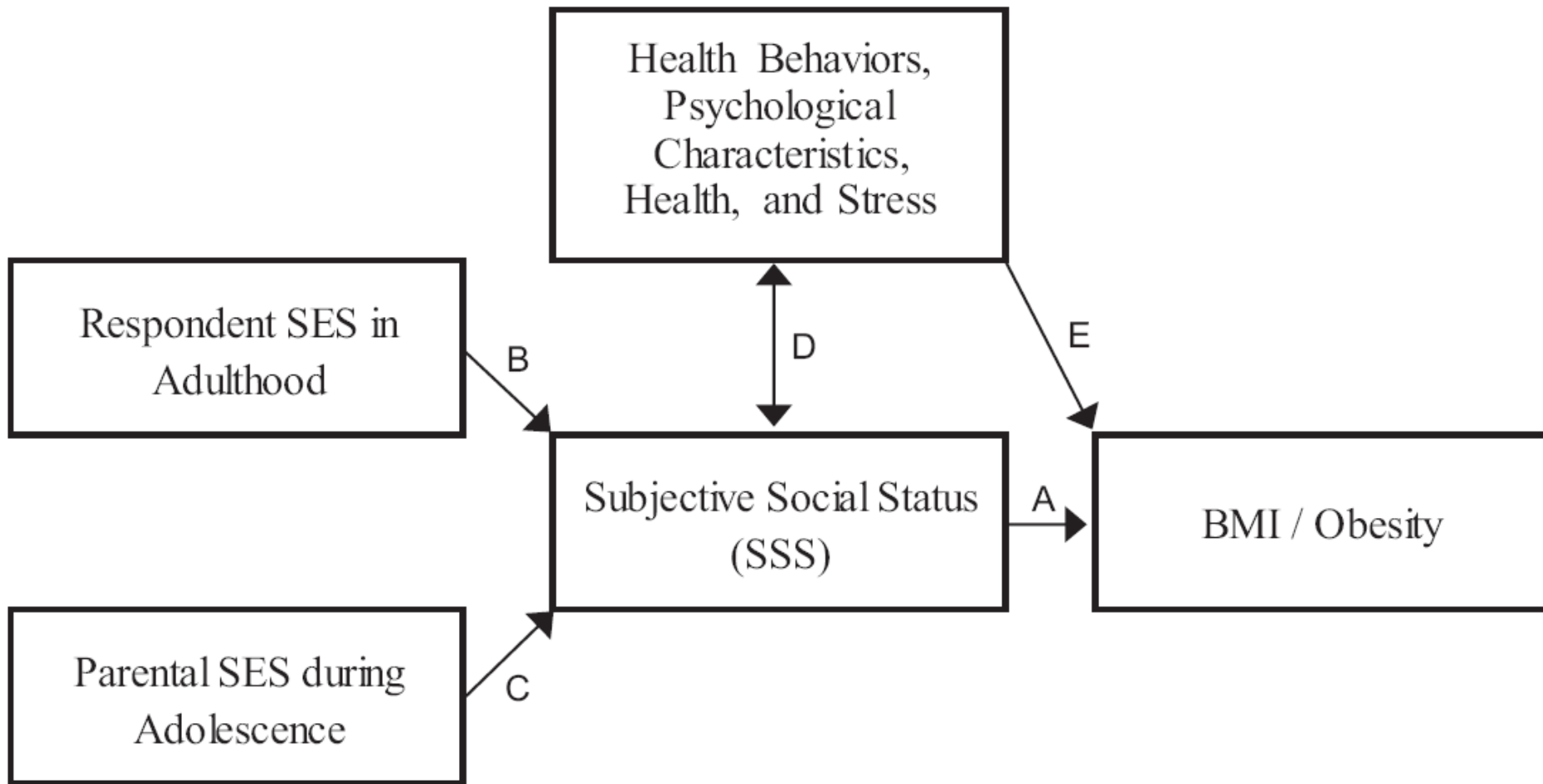
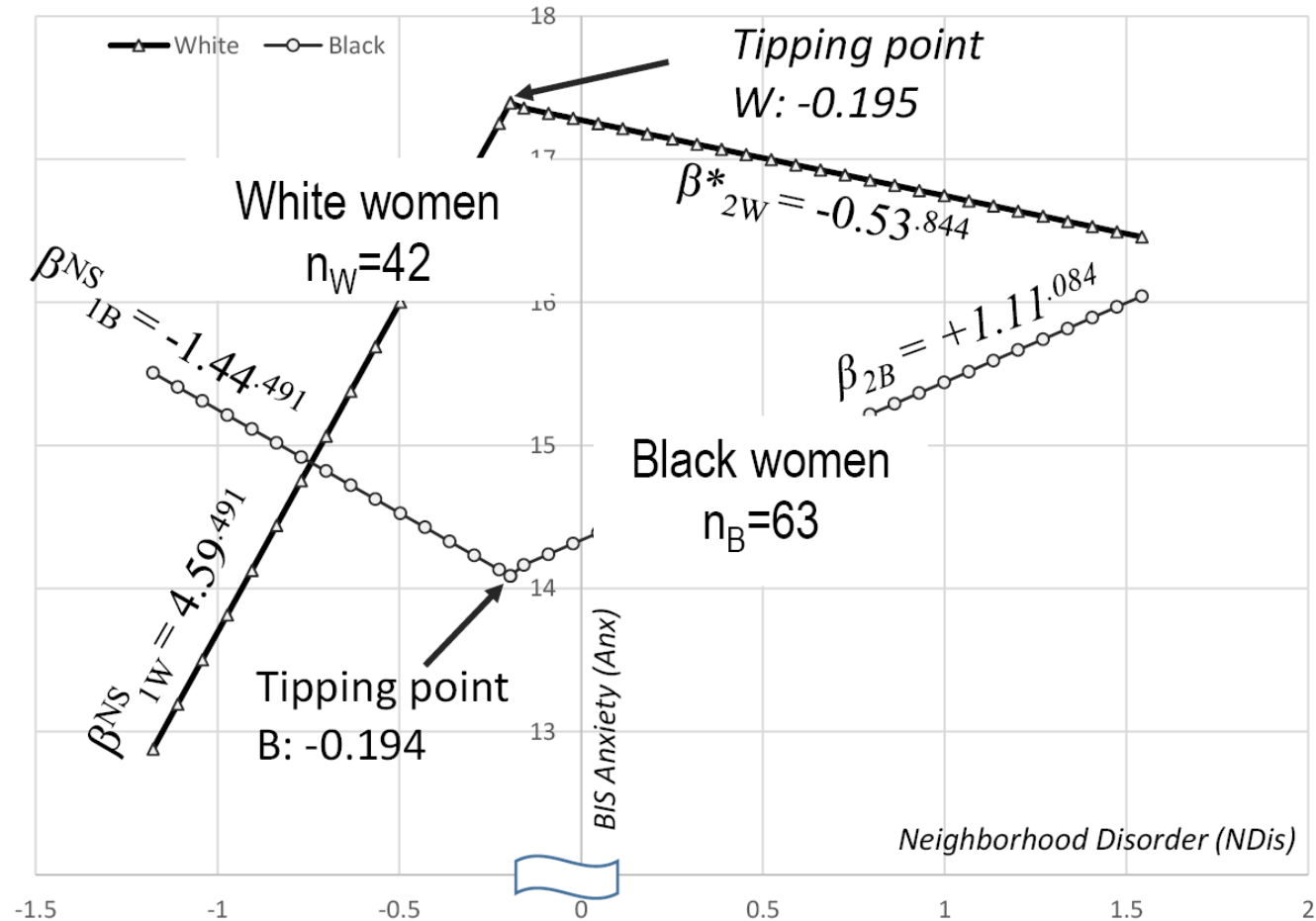


Figure 1. Conceptual model. SES = Socioeconomic status; BMI = Body mass index.

A test of the changing effect

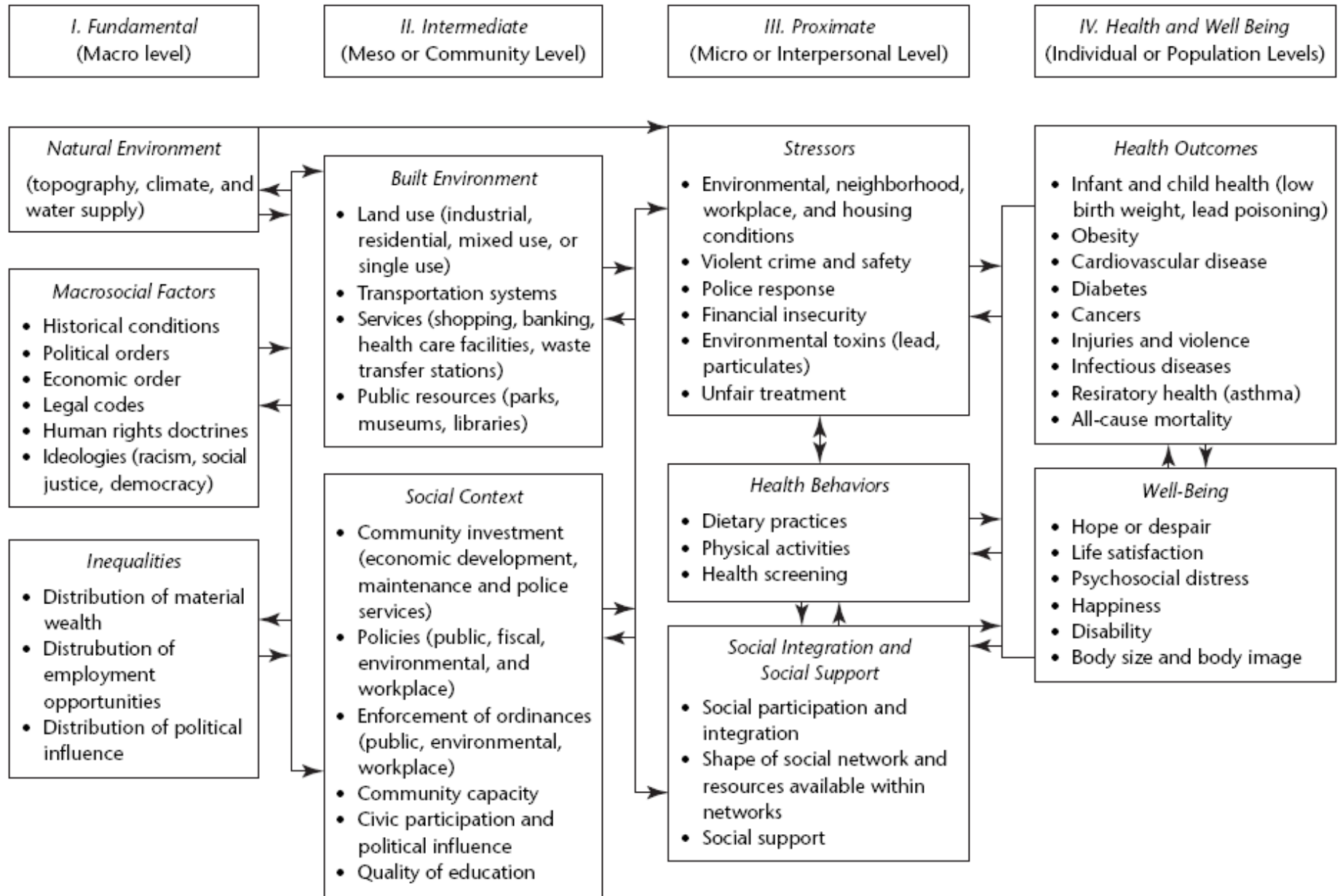
Figure 1. Linear slopes for neighborhood disorder -> anxiety effects, shifting around similar tipping points.

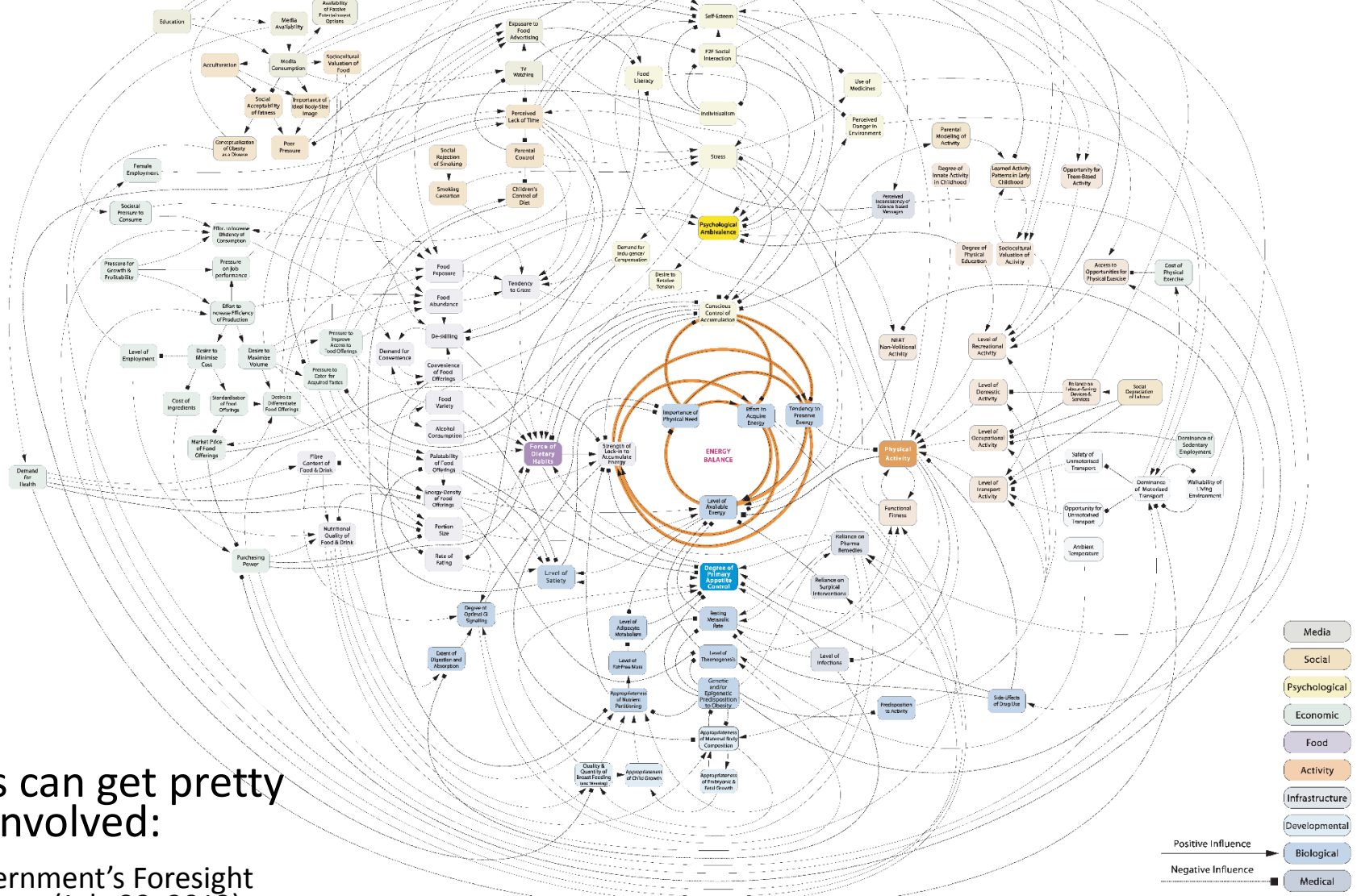
Shifting effects, by race/ethnicity



Coman, E. N., & Wu, H. (2018). Examining Differential Resilience Mechanisms by Comparing 'Tipping Points' of the Effects of Neighborhood Conditions on Anxiety by Race/Ethnicity. *Healthcare*, 6(1), 18.

FIGURE 5.3. Environmental levels and their impact on health.





Models can get pretty involved:

UK Government's Foresight Programm. (July 30, 2013). *Tackling Obesities Future Choices – Building the Obesity System Map*. p. 63

Models:

Logic->Theoretical->Conceptual->Causal

What relates to what in what way

The task is to develop candidate causal models as detailed as possible.

Developing a model requires first an extensive causal search for the 'Big picture': next.

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