Race and causality in health disparities research: time for a necessary paradigm shift

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I have no conflicts of interest to disclose.

Overview

- A little about me
- Motivating Example
- Define "Circular" Slump in Racial/Ethnic Disparities Research
- Brief Introduction to Causal Inference
- Important Considerations for Health Disparities Research

A little about me











Passion for research pertaining to racial/ethnic minorities and health disparities research continued...but hit with a philosophical dilemma

- Started questioning why we always start and end with race
 - We observe a racial/ethnic difference.
 - After adjustment for numerous covariates, difference remains. Further research needed.
- Describe racial/ethnic differences often but identification of effective targets for intervention less frequent.
- Searched the literature for guidance from statisticians and non-statisticians about how we should be more optimally operationalizing race.
 - Needed to fundamentally change my approach (unlearn the approach I had been taught) as it relates to examining race and health in research.

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The researcher subsequently constructs her statistical model (accounting for diet, educational attainment, income, access to healthcare, etc.) and gets some adjusted estimate of the effect of race on hypertension.

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NOT VERY MUCH. BUT WHY NOT???

If you were to ask me what this researcher has learned from the results of her analysis, I'd say...

NOT VERY MUCH. FOCUS OF THIS LECTURE

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- We sample from the population, collect data, conduct an appropriate analysis and conclude:
 - We have sufficient evidence to suggest that the risk of disease differs between Blacks and Whites with magnitude $\geq \Delta$.

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 - "In practice, however, we observe that population distributions of disease vary on the basis of skin color. If we pick a physical characteristic other than race at random, we cannot replicate the same degree of variation in disease occurrence. Blood-type, for example, is easily measured and could provide an alternate classification scheme, but there are few population differences in the distribution of common diseases along this axis. To believe that skin color has a unique association to outcomes ranging from IQ to blood pressure to prostate cancer by sheer chance is a questionable, if not preposterous, proposition."

Kaufman JS, Cooper RS. In search of the hypothesis. Public Health Reports. 1995;110(6):662-666.

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- Templeton (2013) supports Kaufman & Cooper in their efforts to move us away from irrational circular arguments:
 - "Humans are an amazingly diverse species, but this diversity is not due to a finite number of subtypes or races. Rather, the vast majority of human genetic diversity reflects local adaptations and, most of all, our individual uniqueness.

Templeton, A. R. (2013). Biological Races in Humans. *Studies in History and Philosophy of Biological and Biomedical Sciences*, *44*(3), 262–271.

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- Similar to Kaufman & Cooper, Holland (2003) also argues that the "race effect" that we measure does not have any causal interpretation.
 - **Experiences individuals undergo**, not attributes they possess.
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 - Causal variables must reflect the possibility of manipulation.
 - While race is not a causal variable, it can play a crucial role in causal studies for a few reasons:
 - For descriptive reasons
 - "In my opinion, RACE can play an important descriptive role in identifying important societal differences such as those in wealth, education, and health care. The attribution of cause to RACE as the producer of these differences is, to me, the most casual of causal talk and does not lead to useful action."
 - Understanding whether an intervention works differently across racial/ethnic groups
 - Attempting to delve deeper into the effects of important constructs like discrimination and bias

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Ascribing a causal role to race does not get us closer to change.

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As statisticians, mathematicians, data scientists, and clinical investigators, we must scrutinize the treatment of these groups in research too.

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- ▶ But, an association is only a statistical concept.
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 - The statistical relationship will look the same whether $E \rightarrow D$ or $D \rightarrow E$.
 - Also if E and D, have the same underlying cause, we will still observe an association between E and D.
- Causation, however, is not a statistical measure, but stems from a rigorous underlying theory about the relationship between E and D.

- ► To test whether E causes D...
 - If we assume that E and D are both binary, then in theory:
 - We can conceive of a world where every individual has a potential outcome D_E and D_E, : outcome in the presence and absence of the exposure, respectively.
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 - Thus the individual causal effect, $CE_i = D_{E,i} D_{E',i}$.
 - Yet, in reality, we can only observe D_E <u>or</u> D_E.
 - Thus, we make strong assumptions about the comparability of the exposed and unexposed groups, so the unexposed are the counterfactual to the exposed.

- Given this fundamental problem, Holland (1985) suggests we are left with two solutions:
 - Scientific Solution
 - Assume that the response after exposure to the "control" at an earlier time for a given individual is the same as the response after exposure to the "control" for the same individual in the current experiment.
 - Only needs to expose the same individual to the treatment in the current experiment.
 - Underlying conditions may not hold true over time.
 - Difficult to translate to real-world of uncontrolled conditions.
 - Statistical Solution
 - Individual \rightarrow Population
 - Estimate the average causal effect of the treatment/cause over a population
 - Contains individuals who are exposed to the treatment and individuals who are exposed to the control
 - Still requires good study design and that the exposed and unexposed are actually comparable

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 - Reduce bias due to unmeasured confounding.
 - For example, relationship between education and health may be confounded by air quality, Pb exposure, crime, collective efficacy, access to healthy food, etc.
 - Impossible to measure everything w/ limited resources.
 - Instead, study E-D relationship in subpopulations homogenous with respect to a specific contextual construct.

COVID-19 yields increased calls for naming and operationalizing constructs in disparities research

Carrion, D., Colicino, E., Pedretti, N. F., Arfer, K. B., Rush, J., DeFelice, N., & Just, A. C. (2020). Assessing capacity to social distance and neighborhood-level health disparities during the COVID-19 pandemic. *medRxiv*.

Weber, E., Miller, S. J., Astha, V., Janevic, T., & Benn, E. (2020). Characteristics of telehealth users in NYC for COVID-related care during the coronavirus pandemic. *Journal of the American Medical Informatics Association*, *27*(12), 1949-1954.

Nancy Krieger, 2020: ENOUGH: COVID-19, Structural Racism, Police Brutality, Plutocracy, Climate Change—and Time for Health Justice, Democratic Governance, and an Equitable, Sustainable Future. American Journal of Public Health 110, 1620-1623.

Hooper, M. W., Nápoles, A. M., & Pérez-Stable, E. J. (2020). COVID-19 and racial/ethnic disparities. *JAMA*, *323*(24), 2466-2467.

Yancy, C. W. (2020). COVID-19 and African Americans. JAMA, 323(19), 1891-1892.

Almagro, M., & Orane-Hutchinson, A. (2020). JUE insight: The determinants of the differential exposure to COVID-19 in New York city and their evolution over time. *Journal of Urban Economics*, 103293.

Summary

- If we are to move from describing racial differences to identifying mutable targets for intervention, then race CANNOT be our end point.
- Difficult paradigm shift but necessary.
- Baby steps can go a long way.
 - Real World Example: Black-White differences observed in 6-months postpartum depression treatment (prescription medication) acceptability
 - A priori hypothesis that stigma was the culprit did not stand.
 - Medicaid and Black race walked hand in hand.
 - "For example, the vast majority of women in our sample on Medicaid were black and some of our findings related to race may be more reflective of insurance status."

Bodnar-Deren S, Benn EKT, Balbierz A, Howell EA. Stigma and postpartum depression treatment acceptability among black and white women in the first six-months postpartum. *Matern Child Health J*. 2017; doi:10.1007/s10995-017-2263-6.

Race, Genetic Ancestry, and Clinical Algorithms

MEDICINE AND SOCIETY

Embracing Genetic Diversity to Improve Black Health

Akinyemi Oni-Orisan, Pharm.D., Ph.D., Yusuph Mavura, M.S., Yambazi Banda, Ph.D., Timothy A. Thornton, Ph.D., and Ronnie Sebro, M.D., Ph.D.

Article Metrics	March 25, 2021 N Engl J Med 2021; 384:1163-1167 DOI: 10.1056/NEJMms2031080
S RESEARCHERS WHOSE WORK IS LARGELY FOCUSED ON GENETICS and who self-identify as Black men, arguably one of the most disadvantaged groups in the United States, we have had similar formative experiences during our training and careers. We have all dealt with aggressions and microaggressions, isolation, imposter syndrome, the Pygmalion effect or stereotype threat, gaslighting, and a lack of mentorship, especially Black mentorship. We have made our way in a field that has an alarming dearth of leaders and research participants of African descent — a common story among Black professionals in science, technology, engineering, and mathematics (STEM) fields. ¹	Editors Debra Malina, Ph.D., Editor

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We do not believe that ignoring race will reduce health disparities; such an approach is a form of naive "color blindness" that is more likely to perpetuate and potentially exacerbate disparities. Although ignoring race could improve *equality* (by leading to identical medical treatment for everyone), we believe that *equity* is necessary to address disparities. We urge our colleagues in medicine and science to refrain from haphazardly removing race from clinical algorithms and treatment guidelines in response to recent initiatives attempting to combat anti-Black racism. The ultimate goal, we believe, would be to replace race with genetic ancestry in an evidence-based manner. But we have not yet reached a point where genetic-ancestry data are readily available in routine care or where clinicians know what to do with these data. Until we do, ignoring race and thereby reverting to the United States' outdated system of health care, in which clinical research findings are generated in populations of European descent and extrapolated to the treatment of non-European populations, is neither equitable nor safe for those other populations.

QUESTIONS?