Lecture 22: Network scale-up method to study groups most at-risk for HIV

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Sociology 204: Social Networks Princeton University

Pre-read



There are an estimated 38 million people [31.6 million–44.5 million] living with HIV in 2019. In most countries, the disease is concentrated in three high risk groups:

- drug users
- commercial sex workers
- men who have sex with men

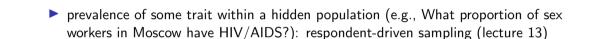
Better information about these group can be used to understand and control the spread of HIV/AIDS: "know your epidemic"



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prevalence of some trait within a hidden population (e.g., What proportion of sex

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network scale-up method (this lecture)

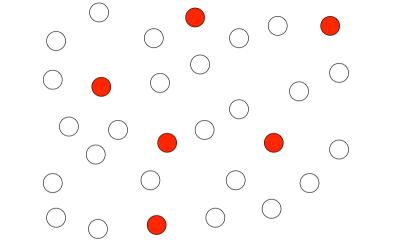
prevalence of some trait within a hidden population (e.g., What proportion of sex workers in Moscow have HIV/AIDS?): respondent-driven sampling (lecture 13)

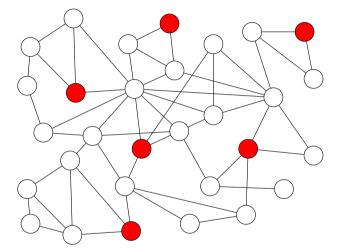
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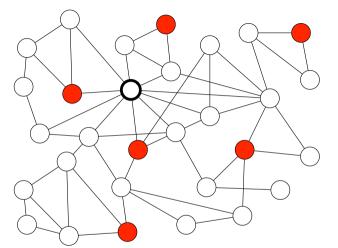
Network scale-up method

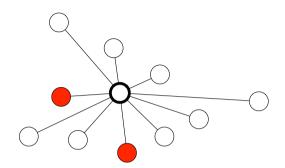


Basic insight from Bernard et al. (1989)





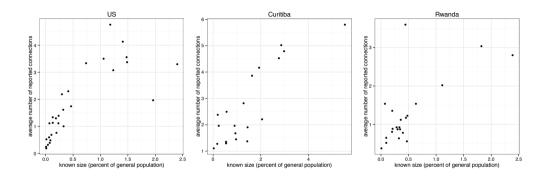




$$\hat{N}_H = \frac{2}{10} \times 30 = 6$$

- ► Requires a random sample from the entire population
- Respondents are asked:
- How many people do you know who are drug injectors?
 - How many women do you know that have given birth in the last 12 months?
 How many people do you know who are middle school teachers?
 - ...
 - ► How many people do you know named Michael?

"Know" typically defined: you know them and they know you and have you been in contact with them over the past two years



On average, these answers are not crazy.

▶ Requires a random sample of the general population, not specific contact with the hard-to-reach population

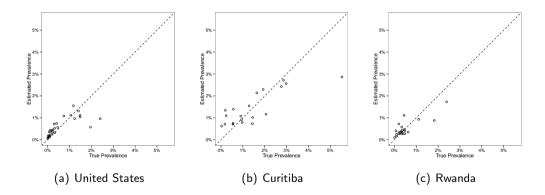
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- Statistical methods are improvable
- ▶ Partially self-validating because it uses groups of known size



Network scale-up method, basic estimator

$$\hat{N}_{H} = \frac{\sum_{i} y_{i,H}}{\sum_{i} \hat{d}_{i}} \times N$$

- \hat{N}_{H} : number of people in the hidden population
- \triangleright $y_{i,H}$: number of people in hidden population known by person i
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$$\hat{d}_i = \frac{2+1}{50.000+1.000} \times 10$$
 million ≈ 600 people

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$$\hat{N}_{H} = \frac{2+4}{400+600} \times 10 \text{ million} = 60,000 \text{ people}$$

GENERALIZING THE NETWORK SCALE-UP METHOD: A NEW ESTIMATOR FOR THE SIZE OF HIDDEN POPULATIONS

Dennis M. Feehan* Matthew J. Salganik[†]

▶ We develop the generalized scale-up estimator and use it to point out possible biases in basic scale-up estimator, but requires two data collections, which is rare (but wait until lecture 23)

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- ▶ We develop the generalized scale-up estimator and use it to point out possible biases in basic scale-up estimator, but requires two data collections, which is rare (but wait until lecture 23)
- ► For the purposes of this class, focus on basic scale-up estimator and the correction factors needed for it

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Practice of Epidemiology

Quantity Versus Quality: A Survey Experiment to Improve the Network Scale-up Method

Dennis M. Feehan®, Aline Umubyeyi, Mary Mahy, Wolfgang Hladik, and Matthew J. Salganik

* Correspondence to Dr. Dennis M. Feehan, Department of Demography, College of Letters and Science, University of California, Berkeley, 2232 Piedmont Avenue, Berkeley, CA 94720 (e-mail: feehan@berkeley.edu).

► Are there tie definitions other than "to know" that would lead to better estimates? We did a survey experiment to compare different definitions and then combined them to produce a single estimate

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- ► Are there tie definitions other than "to know" that would lead to better estimates? We did a survey experiment to compare different definitions and then combined them to produce a single estimate
- Notice how we decided which is better and how we combined estimates
- Notice how we dealt with the many things we did not know with sensitivity analysis.

Enjoy the reading