Statistical estimation of network models from egocentrically sampled network data

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Recent Advances in Statistical Network Analysis

2019 Symposium on Data Science and Statistics

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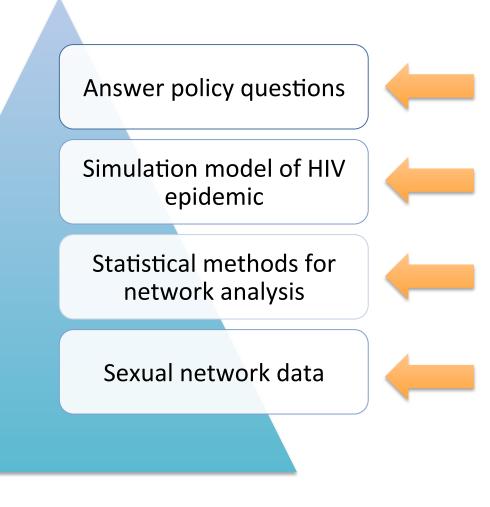
INFERENCE FOR SOCIAL NETWORK MODELS FROM EGOCENTRICALLY SAMPLED DATA, WITH APPLICATION TO UNDERSTANDING PERSISTENT RACIAL DISPARITIES IN HIV PREVALENCE IN THE US

BY PAVEL N. KRIVITSKY^{1,2} AND MARTINA MORRIS¹

University of Wollongong and University of Washington

Egocentric network sampling observes the network of interest from the point of view of a set of sampled actors, who provide information about themselves and anonymized information on their network neighbors. In survey serve certain classes of networks, with the sexual networks that underlie HIV ering some descriptive network features, there is no rigorous and practical statistical foundation for estimation and inference for network models from

Expanding the scope of network analysis for HIV epidemic modeling



Outline

Network analysis in HIV

Sampled sexual network data for HIV modeling

Statistical methods for analyzing network data

- Exponential-family random graph models (ERGMs)
- Extension to sampled data
- Integration with epidemic modeling

Application

Heterosexual HIV dynamics in Seattle/King County

Network analysis in HIV

Network structure matters to HIV

Central to population-level disease transmission, when contacts are

- Rare
- Systematically heterogeneous in probability

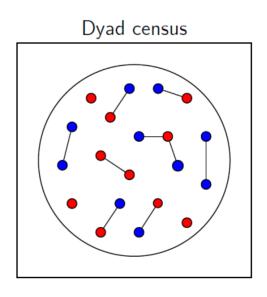
Empirical network research = methods + data

Substantial progress in statistical methods in last 20 years...

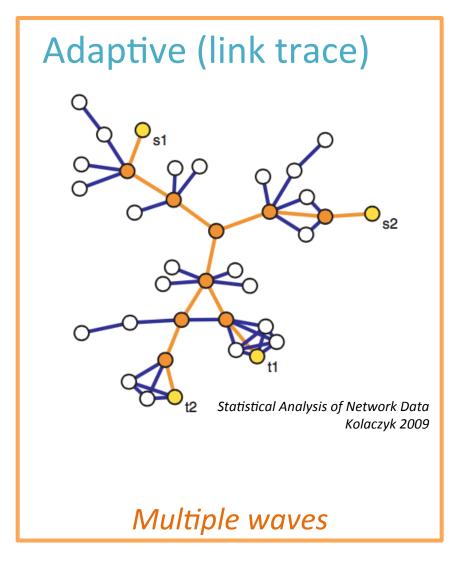
Exponential-family random graph models (ERGMs)

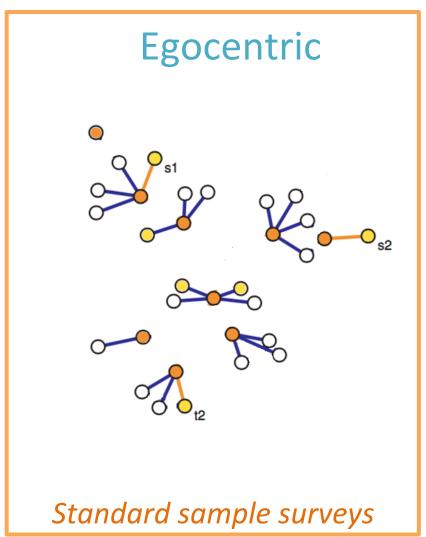
...based on network "census" data

- Often impractical
- e.g. population-level sexual networks

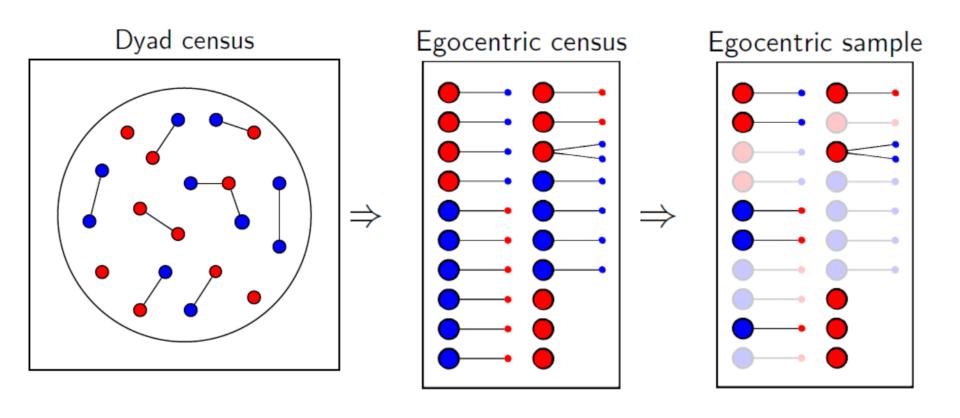


Two types of network sampling





Egocentric sampling



Observe the complete network

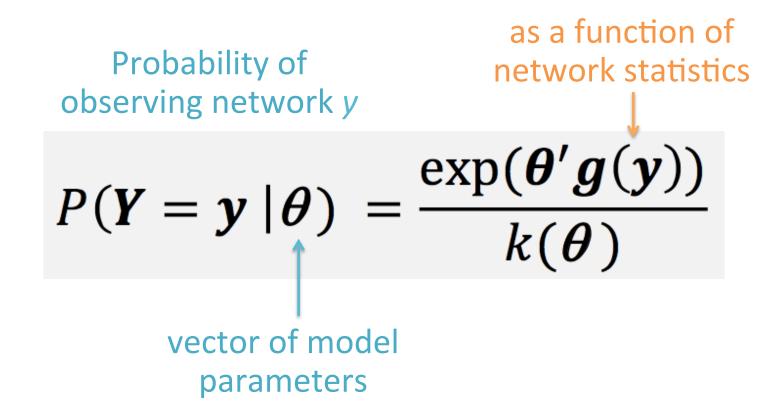
Observe all egos + Reported info on alters

Sample egos + Reported info on alters

Statistical methods for analyzing network data

ERGMs

Foundation: exponential-family random graph models (ERGMs)



Re-expressed at the dyad level

$$logit(P(Y_{ij} = 1 | Y_{ij}^c)) = log \left[\frac{P(Y_{ij} = 1 | Y_{ij}^c)}{P(Y_{ij} = 0 | Y_{ij}^c)} \right]$$
$$= \theta' \delta(g(y))$$
"change statistic"

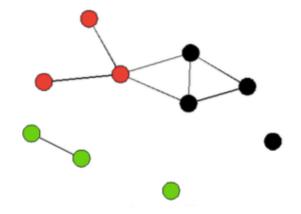
 θ is the per-unit change in the log odds of a tie

Common network statistics g(y)

Edges

Nodefactor/ nodecov

Homophily



7 edges

Red degree = 1 + 1 + 4

Red-Red edges = 2

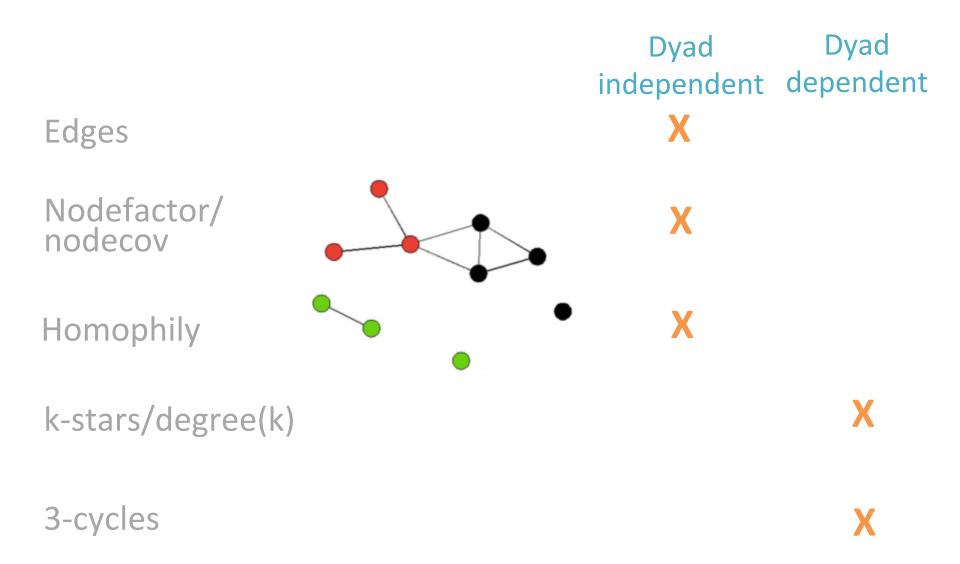
Degree(0) = 2

k-stars/degree(k)

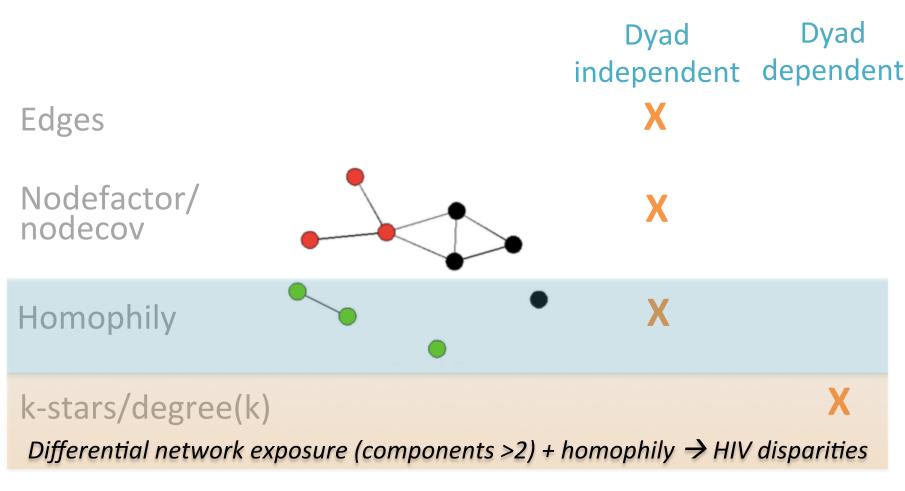
3-cycles

triangles = 2

Common network statistics g(y)



Common network statistics g(y)



ERGM estimation

MCMC Maximum Likelihood Estimation

- For dyad dependent models
- Uses MPLE as initial starting value

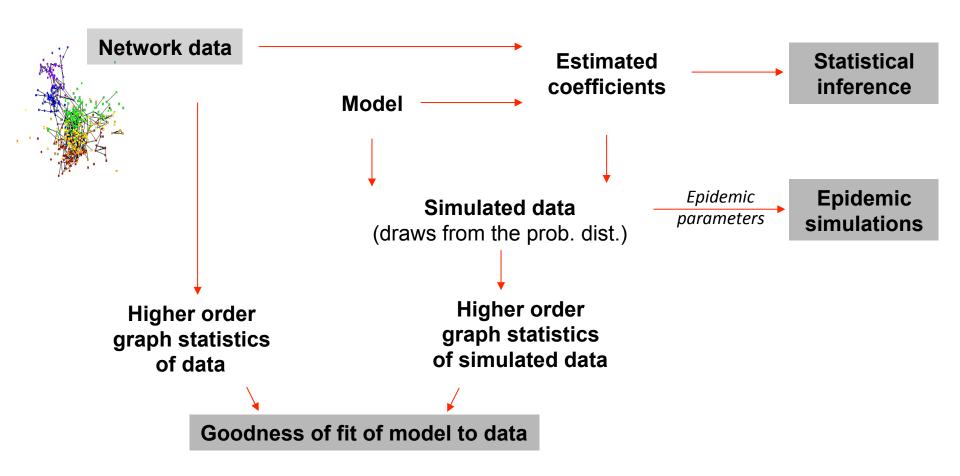
Metropolis Hastings algorithm

- Sampling from the distribution of networks at each iteration
- Given candidate θ_i

Effectively: a network simulation algorithm*

- Proposing toggles, one dyad at a time
- And selecting networks from the chain at suitable intervals

Simulating from a fitted model



Temporal ERGMs

Model link formation and dissolution over time

Krivitsky & Handcock 2014, "A Separable Model for Dynamic Networks"

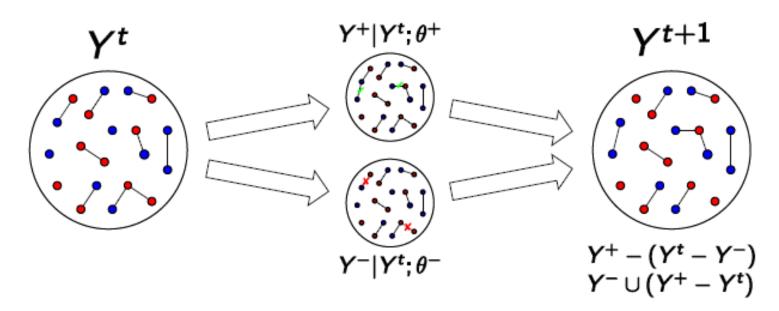


Temporal ERGMs

STERGMs = Separable Temporal ERGMs

- Independent within a time step
- Markov dependent between time steps

a formation ERGM



In R: statnet suite

ergm and tergm

www.statnet.org

statnet

Software tools for the analysis, simulation and visualization of network data.

Welcome to statnet!

Visit the **statnet Wiki** for information on, background material for and access to the **statnet** suite of packages for network analysis. You can find installation instructions, tutorials, and developer resources at the wiki.

What is statnet?

statnet is a suite of software packages for network analysis that implement recent advances in the statistical modeling of networks. The analytic framework is based on Exponential family Random Graph Models (ergm). statnet provides a comprehensive framework for ergm-based network modeling, including tools for model estimation, model evaluation, model-based network simulation, and network visualization. This broad functionality is powered by a central Markov chain Monte Carlo (MCMC) algorithm.

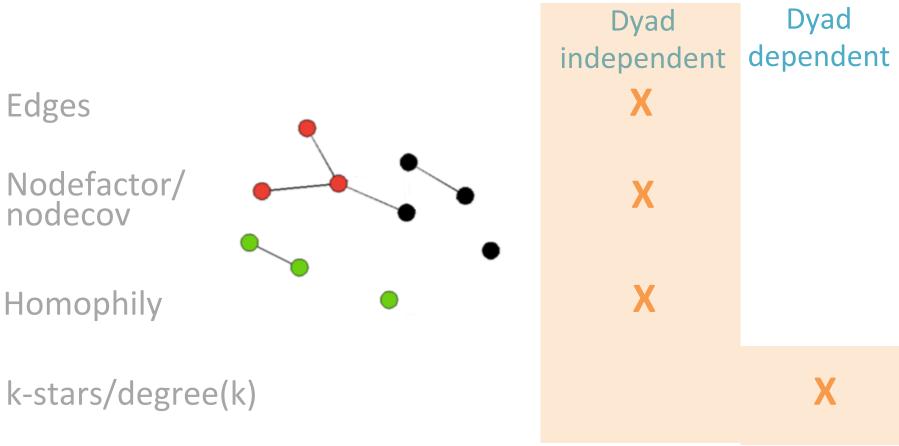
statnet has a different purpose than the excellent packages UCINET or Pajek; the focus is on statistical modeling of network data. The statistical modeling capabilities of statnet include ERGMs, latent space and latent cluster models. The packages are written in a combination of (the open-source statistical language) R and (ANSI standard) C, and are called from the R command line. And because it runs in the R package (www.r-project.org), you also have access to the full functionality of R, including the packages "network" and "sna" written by Carter Butts. statnet has a command line interface, not a GUI, with a syntax that resembles R.

Statistical methods for analyzing network data

Extension to sampled data

Egocentric design determines observable sample statistics

Observable in egocentric sample



3-cycles

X

Key ideas for egocentric estimation

1. Use sample statistics to estimate population statistics g(y)

Requires a scaling assumption

- Assume mean degree is the scale invariant property
- Use inverse probability weighted Hájek estimator of g(y)
- These are the sufficient statistics for estimating the ERGM

- 2. Use estimated population statistics to estimate PMLE of θ (Binder 1983)
 - And variance of this estimate

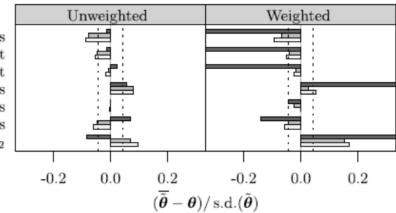
Simulation studies indicate good properties

Krivitsky & Morris 2017

Bias

Sampling weights require larger N to minimize bias

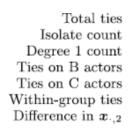
Total ties Isolate count Degree 1 count Ties on B actors Ties on C actors Within-group ties Difference in $x_{\cdot,2}$

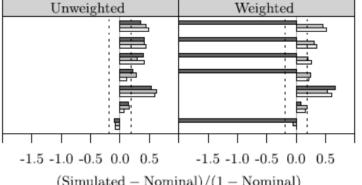




Coverage

Estimated standard errors appear slightly conservative







(Simulated - Nominal)/(1 - Nominal)

In R: ergm.ego

Also part of the statnet suite

www.statnet.org

statnet

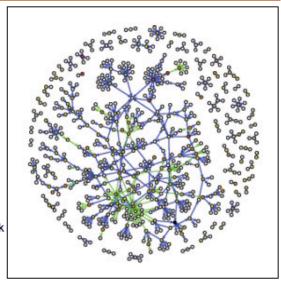
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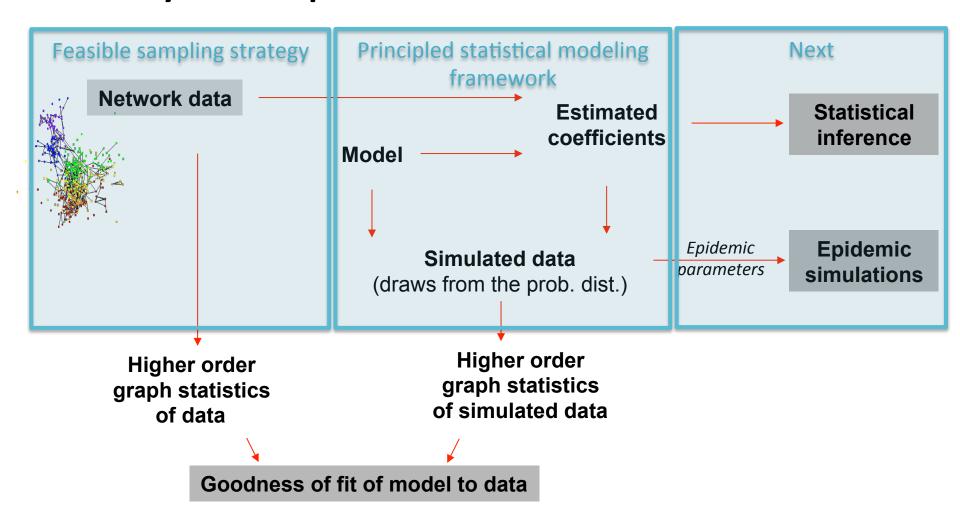
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Ready for epidemic simulations

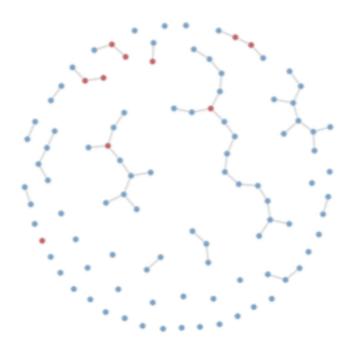


Statistical methods for analyzing network data

Integration with epidemic modeling

HIV epidemic model overview

Foundation = dynamic network (STERGM)



Other processes overlaid

All interact with dynamic network

Demographics

- Sex, age, race structure
- Mortality

Behavior

- Coital frequency
- Condom use

Infectivity by

- Clinical disease stage
- Diagnosis & treatment

In R: *EpiModel*

Integrates disease transmission with STERGMs

www.epimodel.org

EpiModel

Tutorials

Gallery

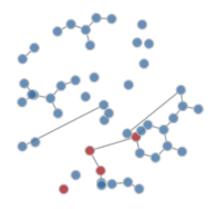
Workshops

Details

EpiModel

Mathematical Modeling of Infectious Disease Dynamics

EpiModel is an R package that provides tools for simulating and analyzing mathematical models of infectious disease dynamics. Supported epidemic model classes include deterministic compartmental models, stochastic individual contact models, and stochastic network models. Disease types include SI, SIR, and SIS epidemics with and without demography, with utilities available for expansion to construct and simulate epidemic models of arbitrary complexity. The network model class is based on the statistical framework of temporal exponential random graph models (ERGMs) implementated in the Statnet suite of software for R.



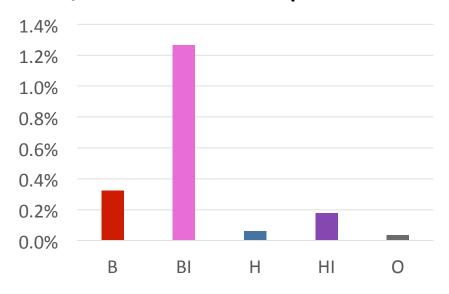
Application to heterosexual networks & HIV in Seattle/King County

Local HIV dynamics & questions

Small heterosexual epidemic

- 10% of new diagnoses
 - Close to eradication?
- Racial disparities in prevalence

S/KC heterosexual HIV prevalence



Local HIV dynamics & questions

- 1. Does the network structure contribute to observed racial disparities?
- Differential network exposure + homophily
- 2. Proof of concept: Can we reproduce the profile of the local epidemic?
- Disparities by race/immigration?
- 3. How close is heterosexual transmission to dying out (the epidemic threshold)?
- Do "bridge" contacts determine the persistence of HIV?

Potential "bridge" contacts

1. Men who have sex with men – via "MSMF"

In the Western US:

 52% of heterosexual female HIV cases phylogenetically linked to MSM (Oster 2015)

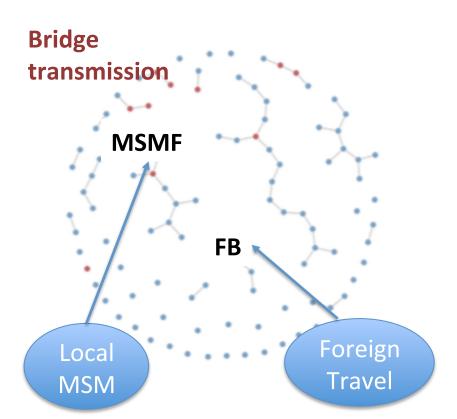
2. Foreign-acquired infections – via "FB"

In Seattle/KC:

- 41% of diagnoses in Blacks among foreign-born
- 52% of diagnoses in Hispanics among foreign-born

Local model structure & data

Foundation = dynamic network (STERGM)



Other processes overlaid

All interact with dynamic network

Demographics

- Sex, age, race structure
- Mortality

Behavior

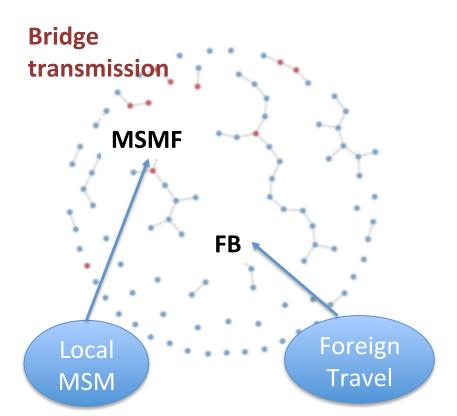
- Coital frequency
- Condom use

Infectivity by

- Clinical disease stage
- Diagnosis & treatment

Local model structure & data

Foundation = dynamic network (STERGM)



Other processes overlaid All interact with dynamic network

Demographics

- Sex, age, race structure
- Mortality

Behavior

- Coital frequency
- Condom use

Infectivity by

- Clinical disease stage
- Diagnosis & treatment

ERGMs on egocentric data

National Survey of Family Growth

- Egocentric survey of most 3 recent sexual partners
 - N = 40,000 respondents (2006-15)
- Weighted to Seattle/King County
 - By age, sex and race/immigration

Three overlapping networks

- Cohabitating
- Persistent
- One-Time

Formation models indicate differential network exposure & homophily

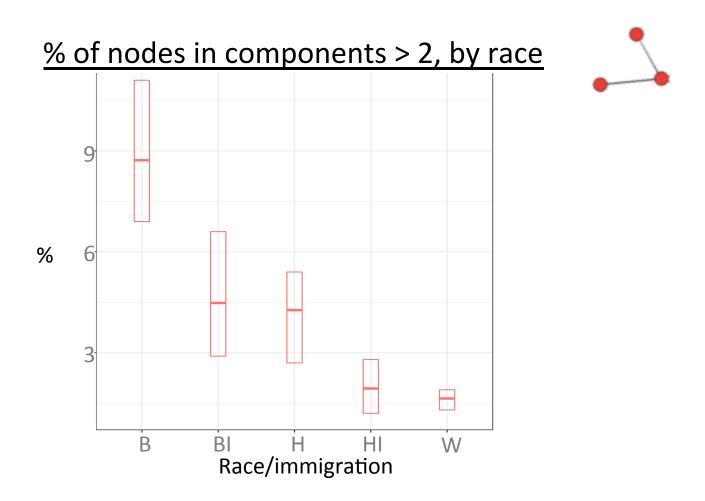
Note:

Actor activity by sex preference MSMF Actor activity by age age age -0.950*** -0.257*** -0.169*** Actor activity by race Black Black Hispanic Hispanic immigrant Actor activity by other-network degree 1 Cohab 1+ Persistent Charle Male w/ 1 Cohab Other female w/ 1 Cohab Other female w/ 1 Cohab Other female Black male Black male Other female Black male Other male Race homophily Black Black Black male Other male Race homophily Black Black manigrant Black		Cohab	Persistent	One-Time
Actor activity by sex preference MSMF Actor activity by age age age Actor activity by age age age aguared Actor activity by race Black Black Black immigrant Hispanic immigrant Actor activity by other-network degree 1 Cohab 1 + Persistent Black female w/ 1 Cohab Other female w/ 1 Cohab Other female w/ 1 Cohab Black female Other female Black male Black male Other female Black male Other male Black male Other male Black B		(1)	(2)	(3)
Actor activity by sex preference MSMF Actor activity by age age age -0.950*** -0.257*** -0.169*** Actor activity by race Black Black Hispanic Hispanic immigrant Actor activity by other-network degree 1 Cohab 1+ Persistent Charle Male w/ 1 Cohab Other female w/ 1 Cohab Other female w/ 1 Cohab Other female Black male Black male Other female Black male Other male Race homophily Black Black Black male Other male Race homophily Black Black manigrant Black	edges	26.809***	6.065***	6.648***
Actor activity by age age	Actor activity by sex preference			
age age squared 0.017*** 0.004*** 0.002*** Actor activity by race Black 1.234*** 1.360*** 0.211*** Black immigrant 1.406*** 1.762*** -0.668** Hispanic 3.388*** 2.112*** 0.440*** Hispanic immigrant 1.692*** 1.219*** -0.691*** Actor activity by other-network degree 1 Cohab	MSMF	-2.111***	-1.050***	0.086
Actor activity by race Black Black Black immigrant Black immi	Actor activity by age			
Actor activity by race Black Black Black immigrant Actor activity by other-network degree 1 Cohab 1+ Persistent Chab Cher female w/ 1 Cohab Cher female w/ 1 Cohab Cher male Black female Cher female Black female Cher female Black male Cher male Cher male Cher male Cher male Black male Cher male Che	age	-0.950***	-0.257***	-0.169***
Black 1.234*** 1.360*** 0.211*** Black immigrant 1.406*** 1.762*** -0.668** Hispanic 3.388*** 2.112*** 0.440*** Hispanic immigrant 1.692*** 1.219*** -0.691*** Actor activity by other-network degree 1 1.219*** -0.691*** Actor activity by other-network degree 1.219*** -0.691*** 1 Cohab -3.143*** -0.557*** Black female w/ 1 Cohab -4.481*** -0.557*** Other female w/ 1 Cohab -4.481*** -0.557*** Degree 1 bias by sex and race 1.106*** -4.752*** Degree 1 bias by sex and race 1.163*** 1.252*** Other female 1.163*** 1.632*** Race homophily 8 3.183*** 3.231*** Black immigrant 3.711*** 2.849*** Hispanic immigrant 2.856*** 2.298*** White 3.109*** 2.172***	age squared	0.017***	0.004***	0.002***
Black immigrant 1.406*** 1.762*** -0.668**	Actor activity by race			
Hispanic immigrant 1.692*** 1.219*** 0.440*** Hispanic immigrant 1.692*** 1.219*** -0.691*** Actor activity by other-network degree 1 Cohab	Black	1.234***	1.360***	0.211**
Hispanic Hispanic immigrant 1.692*** 1.219*** -0.691*** Actor activity by other-network degree 1 Cohab -3.143*** -0.557*** 1+ Persistent -6.250*** -0.557*** Black female w/ 1 Cohab -5.325*** Other female w/ 1 Cohab -4.752*** Degree 1 bias by sex and race Black female 1.106*** Black male 6 1.106*** Other female 1.163*** Black male 6 1.252*** Other male 1.252*** Other male 1.252*** Age mixing 1.252*** Degree 1 bias 3.183*** 3.231*** 3.231*** 4.2849*** 4.2856*** 3.298*** 4.298*** 4.298*** 4.298** 4.298*** 4.298*** 4.298*** 4.298*	Black immigrant	1.406***	1.762***	-0.668**
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Black female w/ 1 Cohab	1 Cohab			-3.143***
Other female w/ 1 Cohab	1+ Persistent	-6.250***		-0.557***
Black male w/ 1 Cohab	Black female w/ 1 Cohab		-4.481***	
Other male w/ 1 Cohab Degree 1 bias by sex and race Black female Other female Black male Other male Other male 1.106*** 1.163*** Black male 1.252*** Other male Race homophily Black Black immigrant Hispanic Hispanic Hispanic immigrant 2.856*** White 3.109*** Age mixing	Other female w/ 1 Cohab		-5.325***	
Degree 1 bias by sex and race 1.106*** Other female 1.163*** Black male 1.252*** Other male 1.632*** Race homophily 3.183*** 3.231*** Black immigrant 3.711*** 2.849*** Hispanic 0.050 0.271** Hispanic immigrant 2.856*** 2.298*** White 3.109*** 2.172***	Black male w/ 1 Cohab		-4.320***	
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Age mixing	Hispanic immigrant	2.856***	2.298***	
	White	3.109***	2.172***	
	Age mixing			
Absolute difference of adjusted sqrt(age) -3.207^{***} -2.600^{***} -2.396^{***}	Absolute difference of adjusted sqrt(age)	-3.207***	-2.600***	-2.396***

*p<0.1; **p<0.05; ***p<0.01

Key finding 1:

Differential network exposure by race confirmed in simulated networks

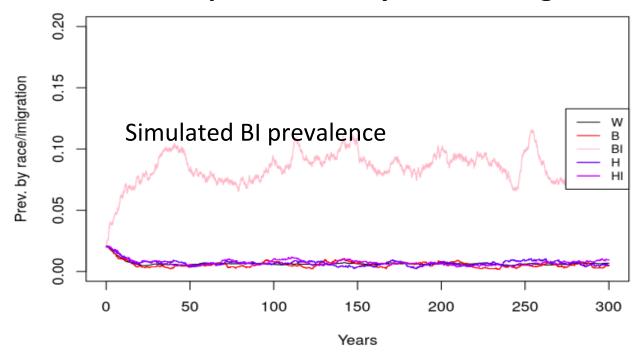


Key finding 2:

Disparity profiles are reproduced

Correct rank ordering of prevalence

Simulated prevalence by race/immigration

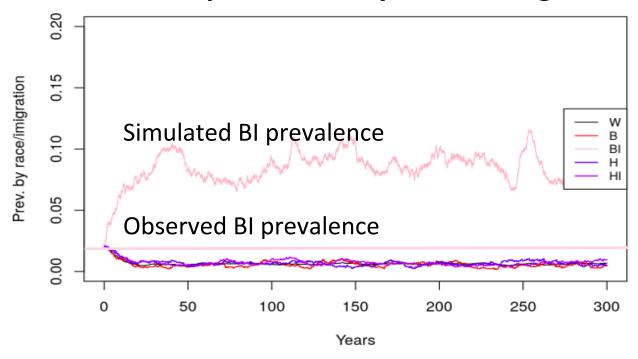


Key finding 2:

Disparity profiles are reproduced

"Bridge" group transmission too high

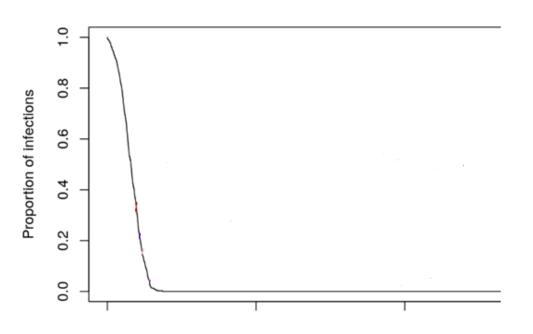
Simulated prevalence by race/immigration



Key finding 3:

Bridging is key to HIV persistence

The epidemic could not be sustained in the local heterosexual population alone



Initial 1% local prevalence

→ epidemic extinction in
~30 years.

Conclusions

ERGM framework + egocentric survey data = feasible network analysis and epidemic modeling

- Test network-structure hypotheses via statistical inference
- Strong basis for statistically-principled epidemic modeling

Supports *local* epidemic modeling

- Model structure based in local features
- Data drawn from local populations
- Investigate locally tailored policies to end HIV

Network Modeling Group

- FACULTY: Martina Morris (UW) Steve Goodreau (UW), Carter Butts (UCI), Mark Handcock (UCLA), Dave Hunter (PSU), Pavel Krivitsky (UNSW), Skye Bender-deMoll (at large), Sam Jenness (Emory)
- UW RESEARCH SCIENTISTS AND STUDENTS: Deven Hamilton
 (Soc), Sara Stansfield (Anthro), Emily Pollock (Anthro), Darcy
 Rao (Epi), Sara Khan (Informatics), Chad Klumb (Math),
 Adam Elder (Biostat)

Resources

statnet

www.statnet.org

EpiModel

www.epimodel.org

"Network Modeling for Epidemics" 5-day course