

# Egocentric ERGM: Application to Persistent Racial Disparities in HIV Incidence

Pavel N. Krivitsky    Martina Morris

Sunbelt Workshop

# Outline

Motivation

Methods

Analysis

# Outline

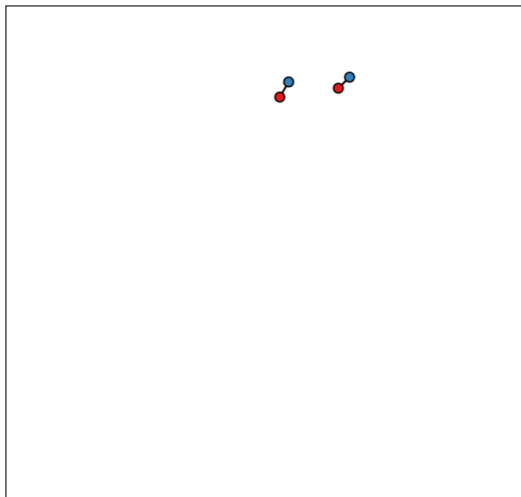
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# Network exposure

Whose behavior matters?



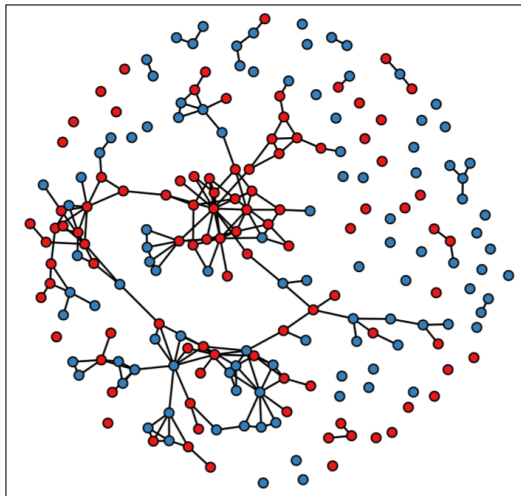
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(Illustration courtesy of Martina Morris (University of Washington). Used with permission.)

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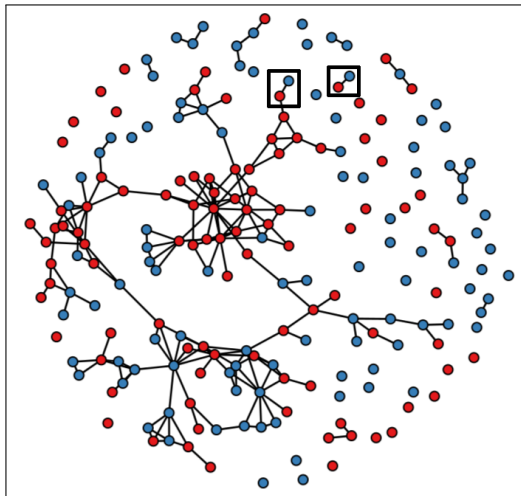
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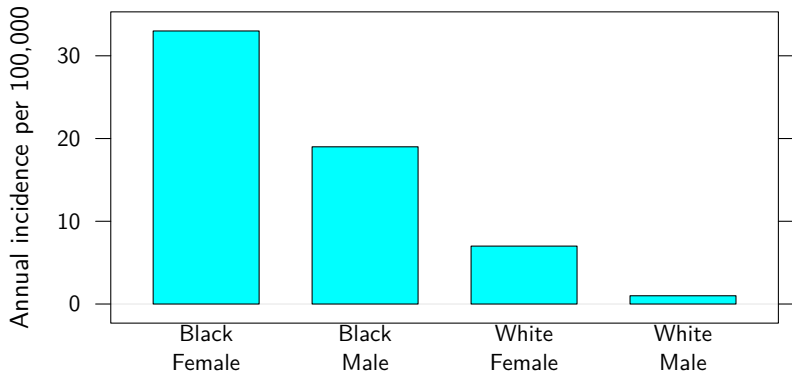
# Racial disparities in HIV prevalence in the US

Morris et al. (2006), Hallfors et al. (2007), and others

- ▶ An African-American is 10 times more likely than a white American to be living with HIV.
- ▶ This has been the case for a long time.
- ▶ Disparity holds across all age and risk groups: heterosexuals, MSM, and drug users.
- ▶ Disparity for African-American women is particularly severe.

# Heterosexually acquired HIV infections in 2010

(National Center for HIV/AIDS, Viral Hepatitis, STD, and TB Prevention (NCHHSTP), 2012)





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- ▶ Other STIs show similar patterns.
  
- ▶ *Individual* behaviors have not been able to fully account for this.
- ▶ No race-linked biological differences have been identified.

# The Network Hypothesis

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3. What impact do these network features have on overall network connectivity and differentials in network exposure by race and sex?

# Not-so-big data

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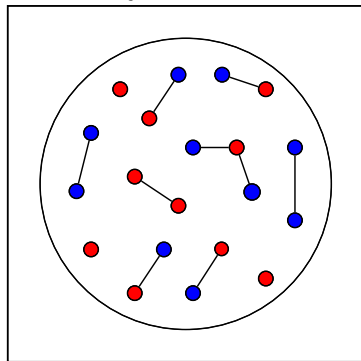
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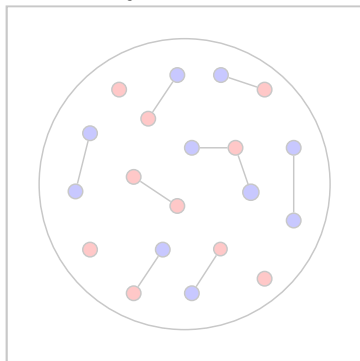
Dyad census



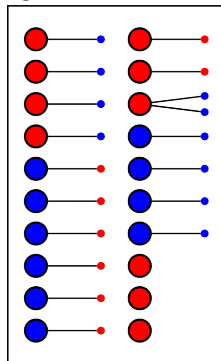
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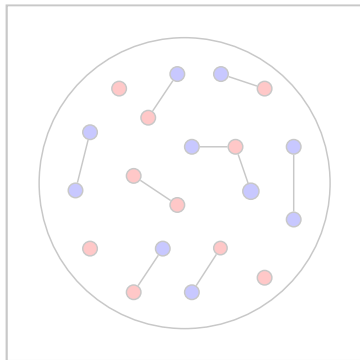
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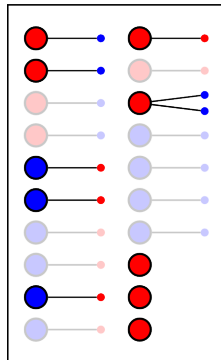
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Egocentric sample



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# Exponential-Family Random Graph Model (ERGM)

(Wasserman and Pattison, 1996; Hunter and Handcock, 2006, and many, many others)

$$\Pr(\mathbf{Y} = \mathbf{y}; \boldsymbol{\theta}, \mathbf{x}) = \frac{\exp\{\boldsymbol{\theta}^\top \mathbf{g}(\mathbf{y}, \mathbf{x})\}}{\kappa(\boldsymbol{\theta}; \mathbf{x})},$$

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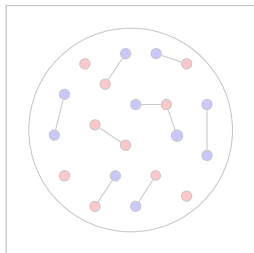
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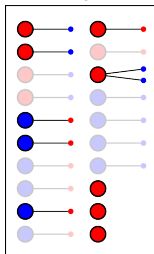
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Or, "Sufficiency to the rescue!"

Dyad  
census

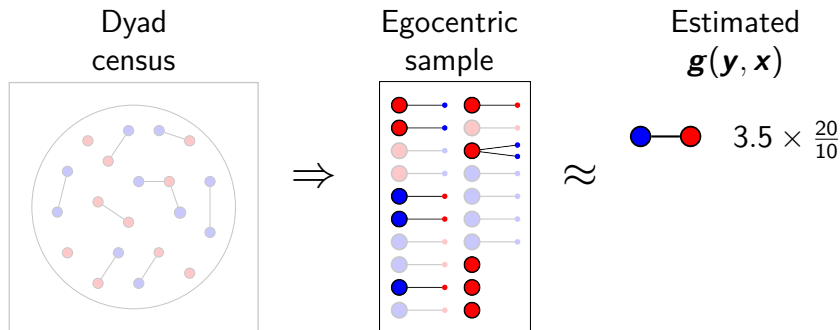


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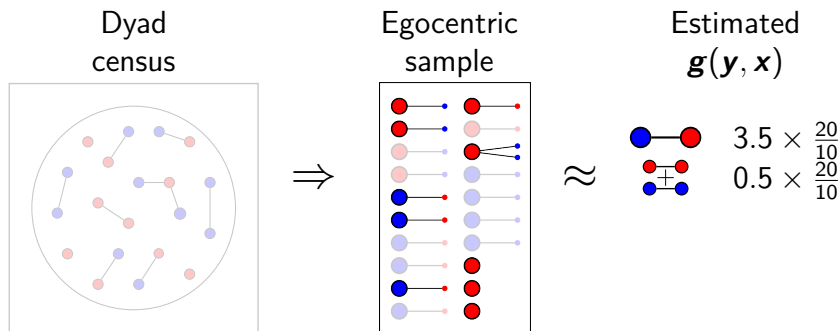
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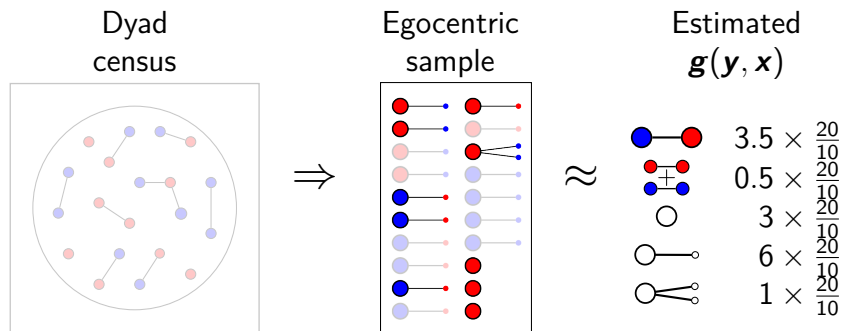
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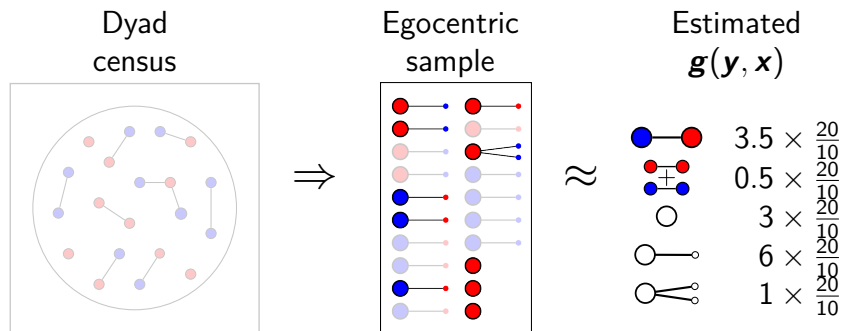
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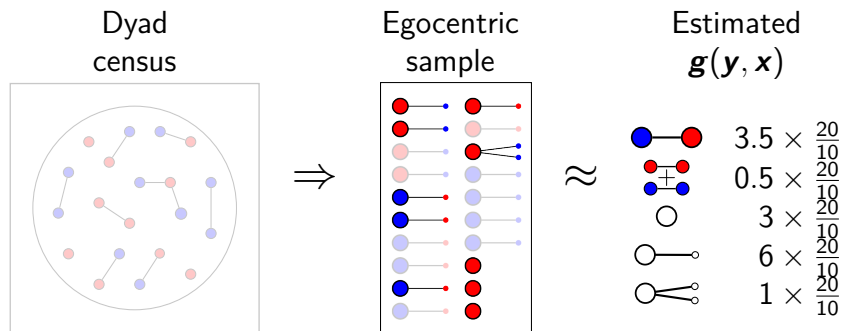
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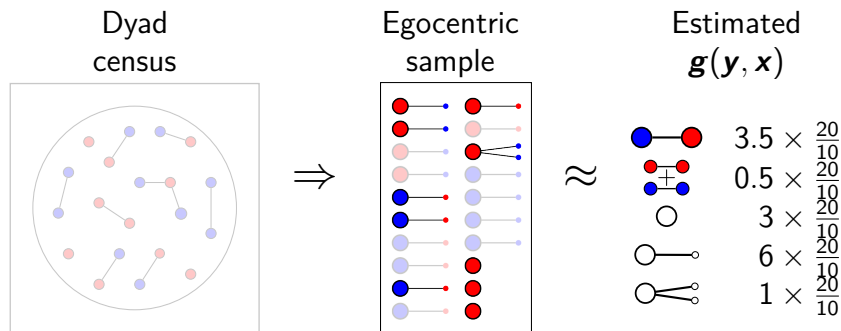
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  - ⇒ Use design-based statistical inference techniques to quantify uncertainty.



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1. How strong is the race homophily in the the population?
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3. What impact do these network features have on overall network connectivity and differentials in network exposure by race and sex?

# 1992 National Health and Social Life Survey (NHSLS) (Laumann et al., 1992, 1994)

- ▶ 3,432 American men and women between 18 and 60 years old
  - ▶ Surveyed age, sex, and race/ethnicity of each respondent and current and some past sexual partner(s)
- ⇒ The data are representative but egocentric.

# The Models

## Model 1 Main Effects

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We will

- ▶ fit these models answer Questions 1 and 2.
- ▶ simulate complete networks from fitted models to answer Question 3.



# Results

After adjusting for network size

<i>Model</i>	1 (Main)	2 (+ Mix.)	3 (+ Monog.)
Actor activity by sex			
Female	0.02 (0.10)	-0.99 (0.19) <sup>***</sup>	-1.88 (0.31) <sup>***</sup>
Male	0.46 (0.10) <sup>***</sup>	-0.55 (0.20) <sup>**</sup>	-1.18 (0.25) <sup>***</sup>
Same-sex	-4.49 (0.21) <sup>***</sup>	-4.50 (0.20) <sup>***</sup>	-4.52 (0.21) <sup>***</sup>
Actor activity by race			
White		0 (baseline)	
Black	-0.09 (0.07)	-0.58 (0.29) <sup>*</sup>	-0.30 (0.38)
Other	-0.03 (0.07)	0.83 (0.33) <sup>*</sup>	0.93 (0.42) <sup>*</sup>
Race homophily by race			
Black		5.13 (0.35) <sup>***</sup>	5.15 (0.38) <sup>***</sup>
Other		2.06 (0.35) <sup>***</sup>	2.04 (0.35) <sup>***</sup>
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Monogamy by sex and race			
Black Female			1.80 (0.47) <sup>***</sup>
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Significance levels: 0.05 $\geq$ * > 0.01 $\geq$ ** > 0.001 $\geq$ ***			

# Results

After adjusting for network size

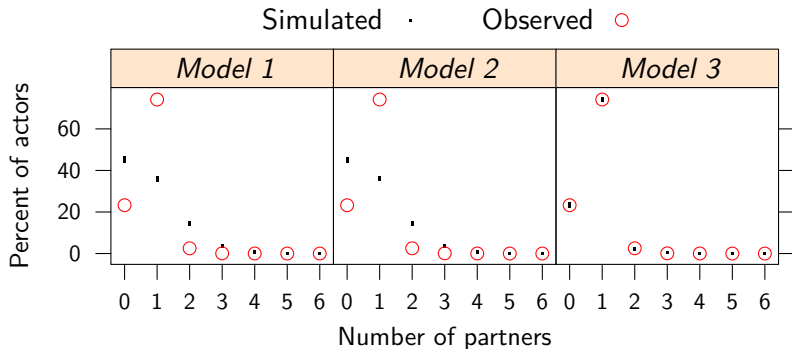
<i>Model</i>	1 (Main)	2 (+ Mix.)	3 (+ Monog.)
Actor activity by sex			
Female	0.02 (0.10)	-0.99 (0.19) <sup>***</sup>	-1.88 (0.31) <sup>***</sup>
Male	0.46 (0.10) <sup>***</sup>	-0.55 (0.20) <sup>**</sup>	-1.18 (0.25) <sup>***</sup>
Same-sex	-4.49 (0.21) <sup>***</sup>	-4.50 (0.20) <sup>***</sup>	-4.52 (0.21) <sup>***</sup>
Actor activity by race			
White		0 (baseline)	
Black	-0.09 (0.07)	-0.58 (0.29) <sup>*</sup>	-0.30 (0.38)
Other	-0.03 (0.07)	0.83 (0.33) <sup>*</sup>	0.93 (0.42) <sup>*</sup>
Race homophily by race			
Black		5.13 (0.35) <sup>***</sup>	5.15 (0.38) <sup>***</sup>
Other		2.06 (0.35) <sup>***</sup>	2.04 (0.35) <sup>***</sup>
White		2.25 (0.34) <sup>***</sup>	2.32 (0.36) <sup>***</sup>
Monogamy by sex and race			
Black Female			1.80 (0.47) <sup>***</sup>
Other Female			2.51 (0.67) <sup>***</sup>
White Female			2.25 (0.31) <sup>***</sup>
Black Male			0.99 (0.24) <sup>***</sup>
Other Male			1.40 (0.31) <sup>***</sup>
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Significance levels: 0.05 $\geq$ * > 0.01 $\geq$ ** > 0.001 $\geq$ ***			

# Results

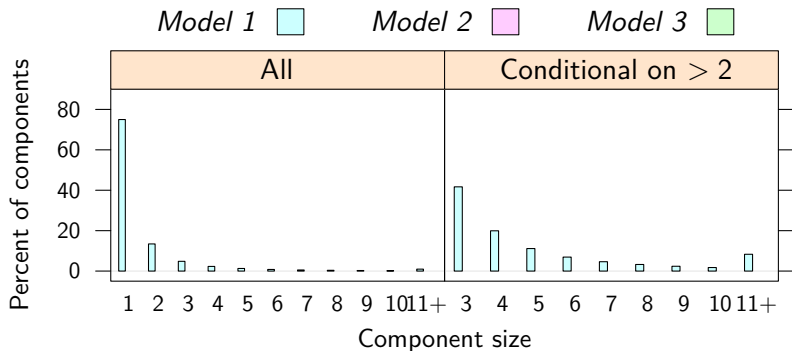
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# Goodness-of-fit of degree distributions

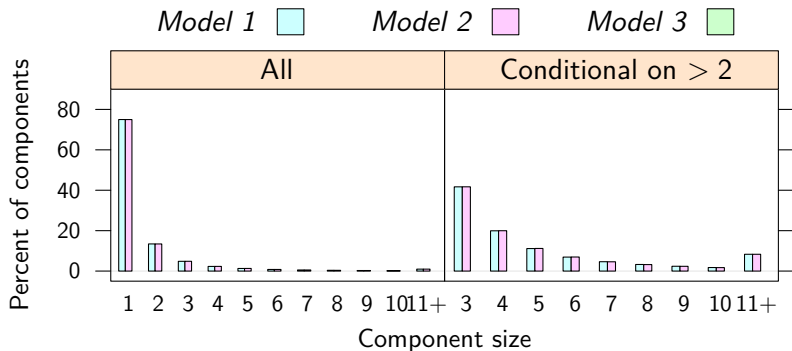


# Simulated distribution of connected component sizes

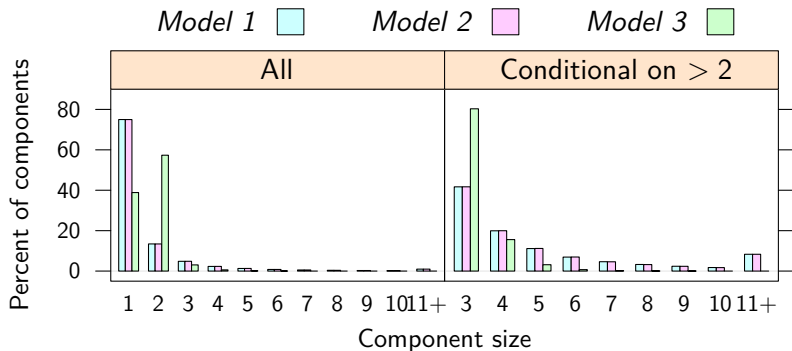




# Simulated distribution of connected component sizes



# Simulated distribution of connected component sizes



# Contrasts of interest

Differences in monogamy:

White Female and Black Female:  $2.25 - 1.80 = +0.45$ ,  
s.e. = 0.51,  $P$ -value  $> 0.3$

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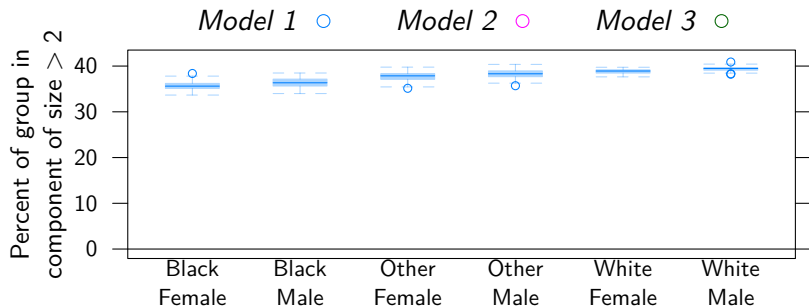
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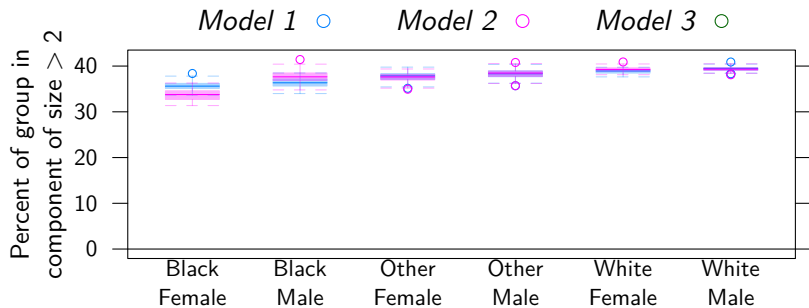
**White Male and Black Male:**  $2.16 - 0.99 = +1.17$ ,  
s.e. = 0.32,  $P$ -value  $< 0.001$

**Female and Male:** Women have higher monogamy coefficients in all groups, but not statistically significant.

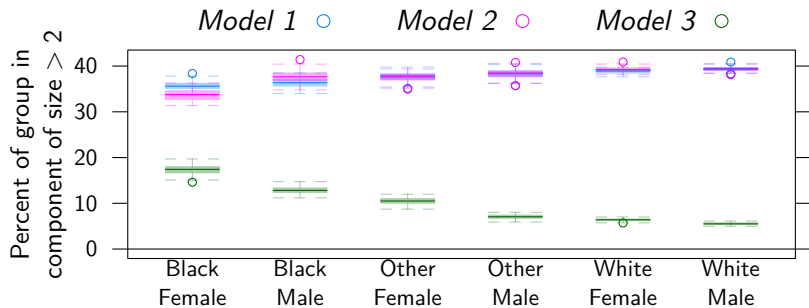
# Simulated network exposure by race and sex



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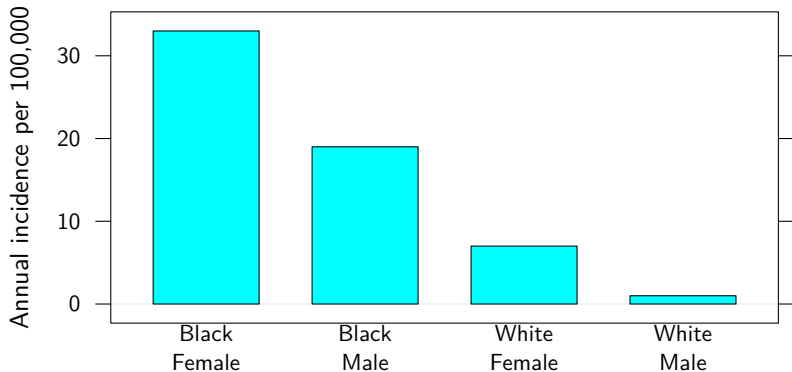


# Simulated network exposure by race and sex

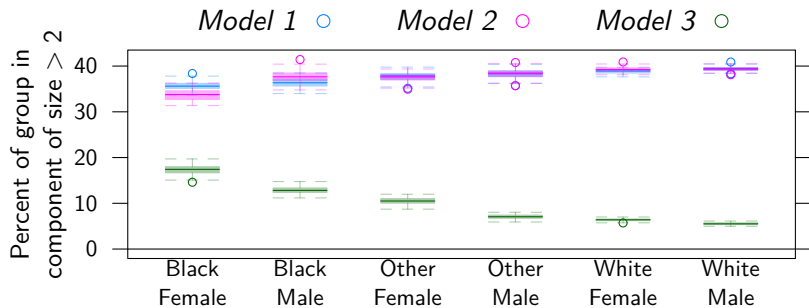


# Heterosexually acquired HIV infections in 2010

(National Center for HIV/AIDS, Viral Hepatitis, STD, and TB Prevention (NCHHSTP), 2012)



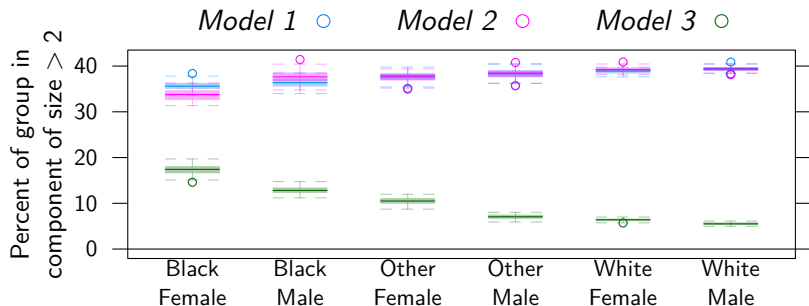
# Simulated network exposure by race and sex



- ▶ Model 3 exposure consistent with observed prevalence of HIV.

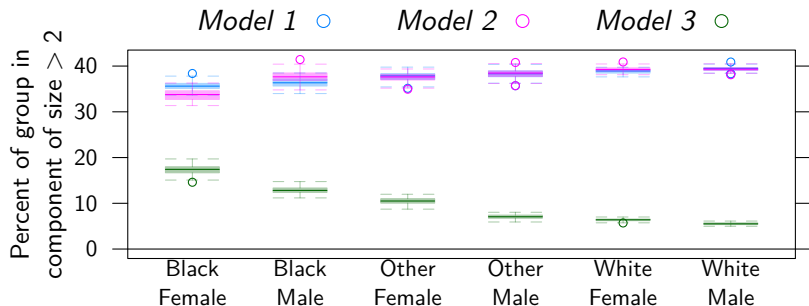


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- ⇒ In a network, what you do might be less important than what those you are connected to do.

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- ▶ Given very limited data, we fit full-network models, and reconstruct networks consistent with the data.

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Tech report available at `https:`

`//niasra.uow.edu.au/workingpapers/UOW190187.html.`

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