Investigating the Impact of an HIV Combination Prevention Strategy in Botswana

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> > Joint work with

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# Goal: to evaluate the impact and cost effectiveness of combination HIV prevention interventions on population-level HIV incidence in Botswana



Map of 50 Botswama communities with population sizes ranging from 3,000 to 15,000 (provided by Hermann Bussmann and colleagues).

### **Design Overview**



- Combination Prevention includes
  - Enhanced HIV testing and counseling (HTC),
  - Prevention of mother-to-child transmission (PMTCT),
  - Enhanced test-linked care(TLC) in relation to ART
  - Male circumcision (MC)
- Both arms will evaluate HIV incidence, from cohorts (20% of the population) followed longitudinally for 3 years.

### **Simulation studies**

Model intervention impact on HIV spread over 4 years

- Generate sexual networks then propagate disease spread on these networks
- Community characteristics:
  - Sexual network characteristics (including mixing between communities)
  - Varying coverage level for different prevention modalities
  - Population sizes
- Individual characteristics:
  - Transmission risk
  - Disease progression
  - Condom use
  - Linkage to care
  - Circumcision status



### Network Construction for a pair of communities A and B



### Mixing matrix allowed to vary

	А	B
Α	40%	10%
B	10%	40%



## **Network Construction**



Bipartite Graph (Relationship only between genders)

- Two Arms (Control and Treatment)
- Control for mixing between the two arms
- Degree (number of partners) Distribution based on Likoma Island

# **Degree Mixing Matrix**



Gender			Male	
	Degree	1	2	3
Female	1	1	0	0
	2	1	2	3

Assortative: monogamous couples; partners who each of whom has many others Dissassortative: married people with sex worker partners

### **Metropolis Hastings Algorithm**

### Edge switching (or toggling) is used to propose networks



Target: P(G=g) is proportional to  $1/(\# \text{ in } C(g)) \cdot P(C(g))$ 

where C(g) is the congruency class, i.e. all networks share a given feature

# MH will accept a proposal network, $p_{t+1}$ . with probability:

 $P(accept) = P(\overline{G} = p_{t+1}) / P(G = g_t)$ 

 $= \frac{\text{\# of graphs in } C(g_t)}{\text{\# of graphs in } C(p_{t+1})} \frac{P(C(p_{t+1}))}{P(C(g_t))}$ 

$$= \frac{f(C(g), C(p_{t+1})) P(C(p_{t+1}))}{f(C(p_{t+1}), C(g)) P(C(g_t))}$$

where f (C(a), C(b)) is the average number of elements in C(a) that are valid proposals from an element in C(b). The value of  $f(\cdot, \cdot)$  can be approximated from the degree distributions (egocentric data) or mixing matrices (network level data).

### Cumulative sexual contact network



# **Network Construction**

### (from cumulative contact network to dynamic network)

**Duration of Relationships** (Estimated from Mochudi data) 0.0015 0.0010 Density 0.0005 0.0000.0 0 2000 4000 6000 8000 10000 Number of Days

Relationship	Duration	Date	Start/End
e1	d1	t1	1
e2	d2	t2	0
e <sub>n</sub>	d <sub>n</sub>	t <sub>n</sub>	1
•	••		

0 4 years

# **Initiation Conditions**

- HIV Prevalence, 24.8%
- Condom use: 40%
- % of males circumcised: 12.7%
- VL/CD4 distribution: based on data from Mochudi
- % on ART among eligible subjects (CD4<350): 60.9%

## **Epidemic Spread**

- Disease progression of HIV infected subjects
  - Duration of high VL after infection: Incidence cohort
  - Rate of CD4 decline: Incidence cohort/Mochudi
- Transmission risk
  - Depending on VL category: Quinn et al. (2000)
  - Reduction in transmission risk
    - From knowledge of serostatus: 30%
    - From condo use: 85%
  - Reduction in acquisition risk from circumcision: 60%

# Mixing

- In Mochudi about 30% of partners of villagers were reported to living outside of the village
- Partnerships do not attenuate treatment effects if partners are from:
  1) different SOC villages
  2) different INT villages
  3) SOC and non-study villages

### Impact of Mixing on Expected Incidence Rates

INT SOC

- 10% mixing: 2.7% vs 5.2%
- 20% mixing: 3.1% vs 5.2%
- 30% mixing 3.2% vs 4.3%
- 40% mixing 3.5% vs 4.6%

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Cumulative Year 3

### Mixing by Activity and Incidence (SOC)

### Sensitive Analysis: Power for Varying Reductions in HIV Incidence in Intervention vs. SOC Arms



For the planned, assuming a k of 0.26, if the cumulative incidence in the intervention arm is  $\geq$  35% lower than that in the SOC arm (3.9%), the study will have >80% power.

# **DSMB** review

- Reviews DSMB 0-4 take place at baseline and the ends of yrs 1, 2, 3 and 4.
- DSMB 1-2 evaluate intervention rollout and safety.
- DSMB 3 can recommend stopping for efficacy or futility, or extending the study by 1 year if:
  - poor intervention coverage, or
  - low incidence in the SOC arm.
- Conditional power assessed by simulation studies.

# **Phylogenetic analysis**

- Analyze DNA or amino acid sequences to gain information about evolutionary relationships
- Can use properties of phylogenetic trees to create test statistics to determine whether certain demographics are more likely to be clustered
- Common algorithms for tree construction:
  - maximum likelihood
  - neighbor joining
  - maximum parsimony



# Conclusions

- Network models can help in designing and monitoring studies of HIV prevention in Africa and in investigating the attenuation resulting from mixing across communities.
- Mixing by degree does not appear to impact the efficacy of interventions.

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# References

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