

# Using mathematical models for health economic analyses

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# Outline

- Introduction to modeling
- Infectious disease modeling
  - Introduction;  $R_0$
- How models can be used to estimate health outcomes
  - Example: Potential impact of ART for prevention
  - What study data can you use to parameterize models
- When to use which model

# An introduction to Mathematical Models

- Framework for understanding and communicating infectious disease\*
- Explicit assumptions help delineate which parameters are based on evidence
- Quantitative or qualitative results are compared with observed or experimental data
- Validated models can be used to estimate the potential impact of interventions (e.g. ART for prevention) on health outcomes
  - HIV incidence cases
  - HIV associated death
  - HIV associated disability adjusted life years (DALYs)

\*Garnett, G. P. (2002). Sex Transm Infect 78(1): 7-12.

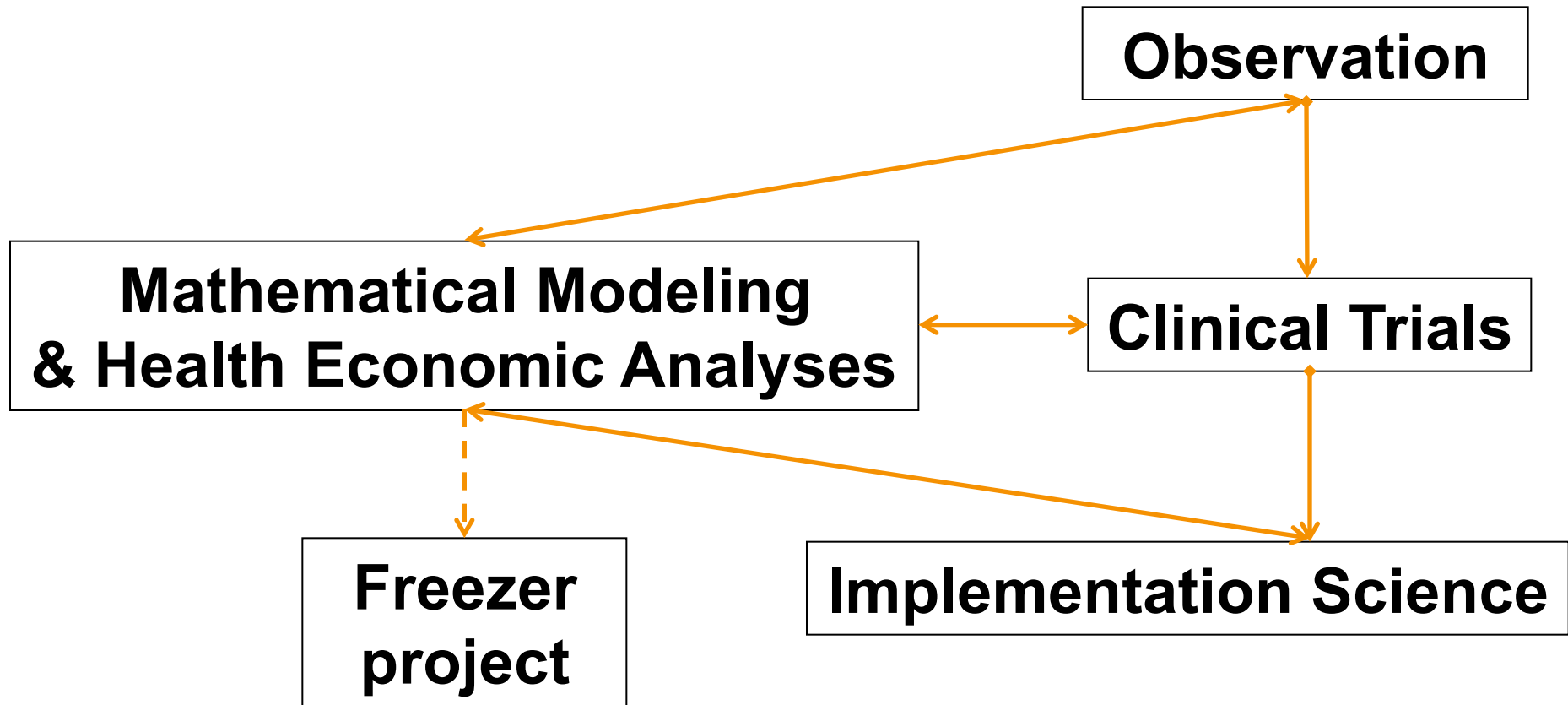
# Models in health economic analyses

- Used to structure the economic question and compare all relevant alternatives
- Extrapolate beyond observed data
- Link intermediate and final endpoints
- Generalize results to other settings/patient groups
- Synthesize evidence to simulate comparisons where RCTs don't exist
- Indicate the need for further research

# Types of models

- Static models – equilibrium (time-invariant)
- Dynamic models – time dependent change
  - Force of infection can change over time
  - Includes herd immunity
- Both static and dynamic models can be either deterministic or stochastic (constrained random variables)
- Choice of model depends on scientific question

# Where do models fit in the path from discovery to implementation?



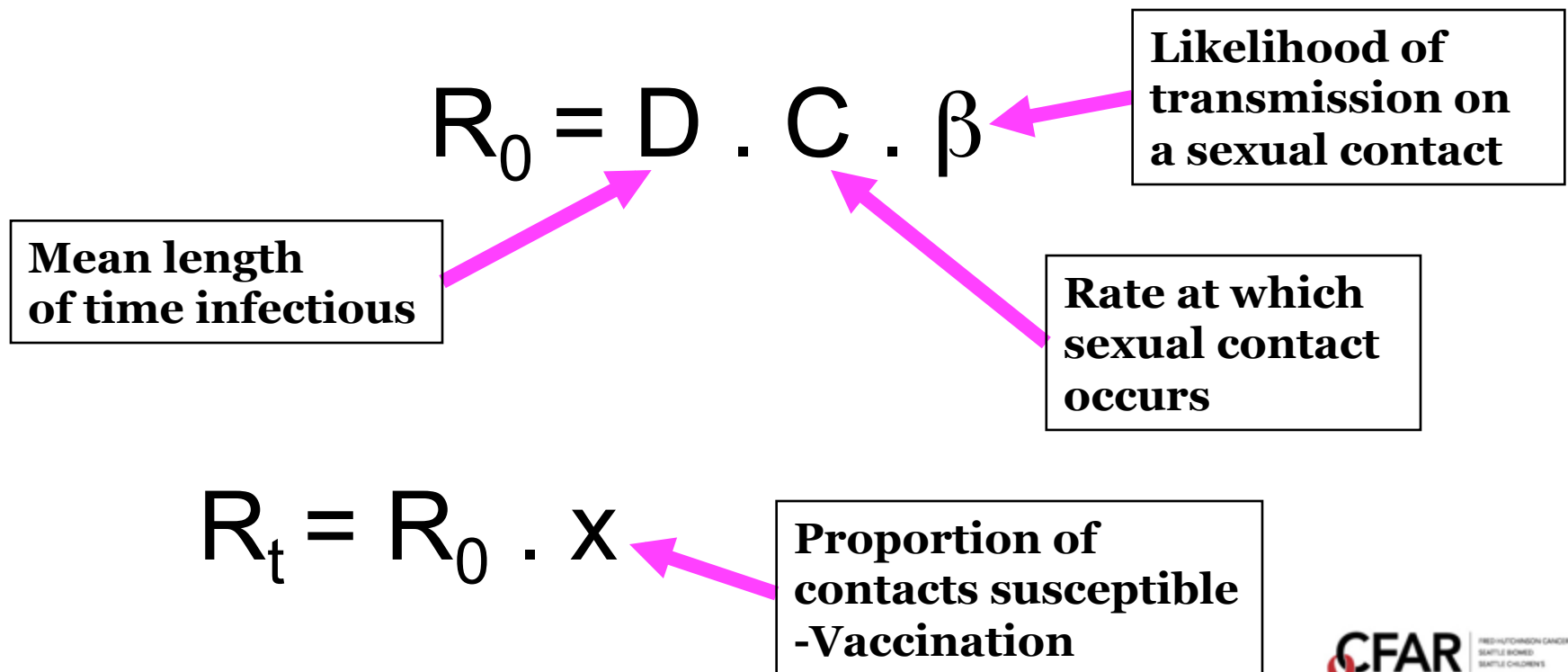
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# The basic and effective reproductive numbers

**$R_0$  The Basic Reproductive Number** - The number of new infections caused by one infection in an entirely susceptible population

**$R_t$  The Effective Reproductive Number** - The number of new infections caused by one infection at a given time

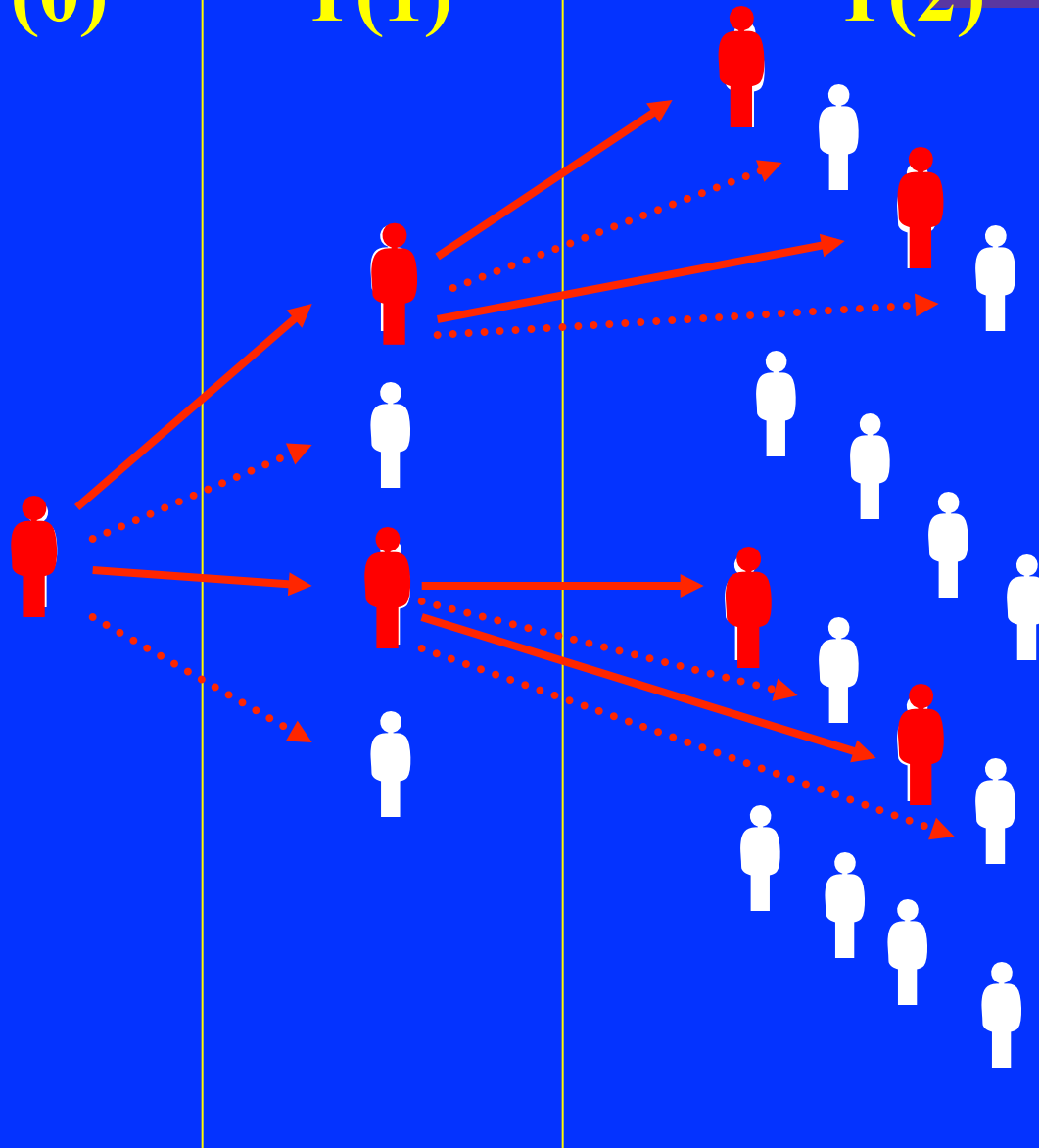




T(0)

T(1)

T(2)

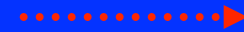


$$R_0 = 2$$

Transmission



No Transmission



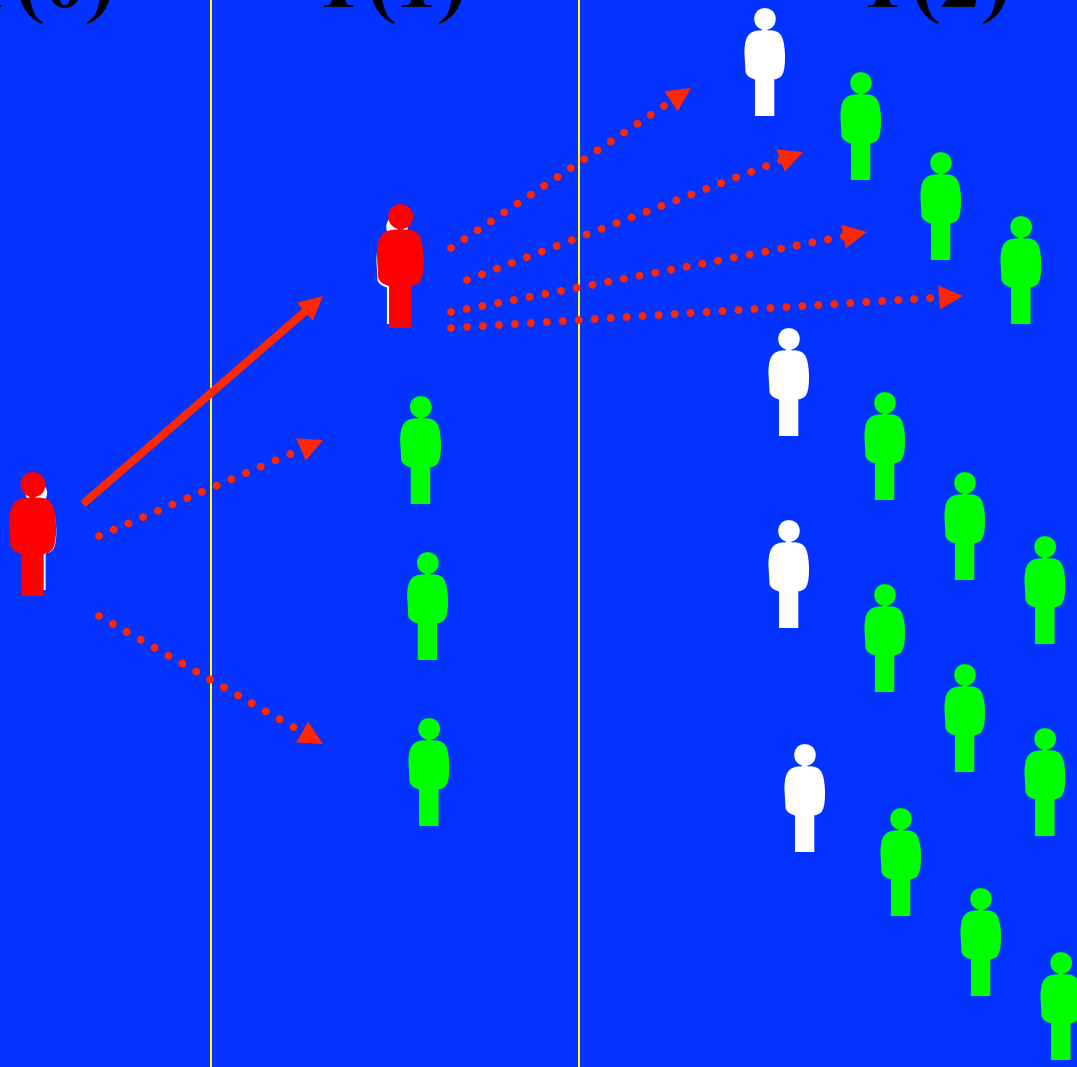
 Infectious

 Susceptible

**T(0)**

**T(1)**

**T(2)**



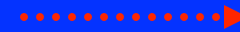
$R_0 = 2$

$R_t = R_0 \cdot \text{prop susceptible}$   
 $= 0.5$

**Transmission**



**No Transmission**



 **Infectious**

 **Susceptible**

 **Immune**

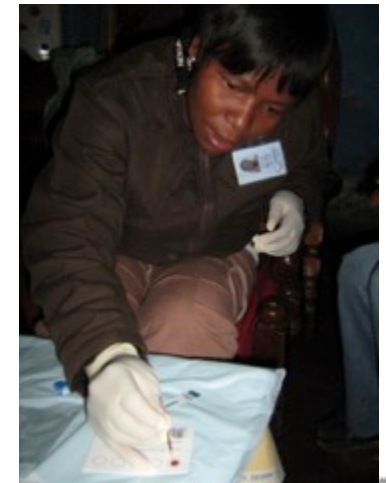
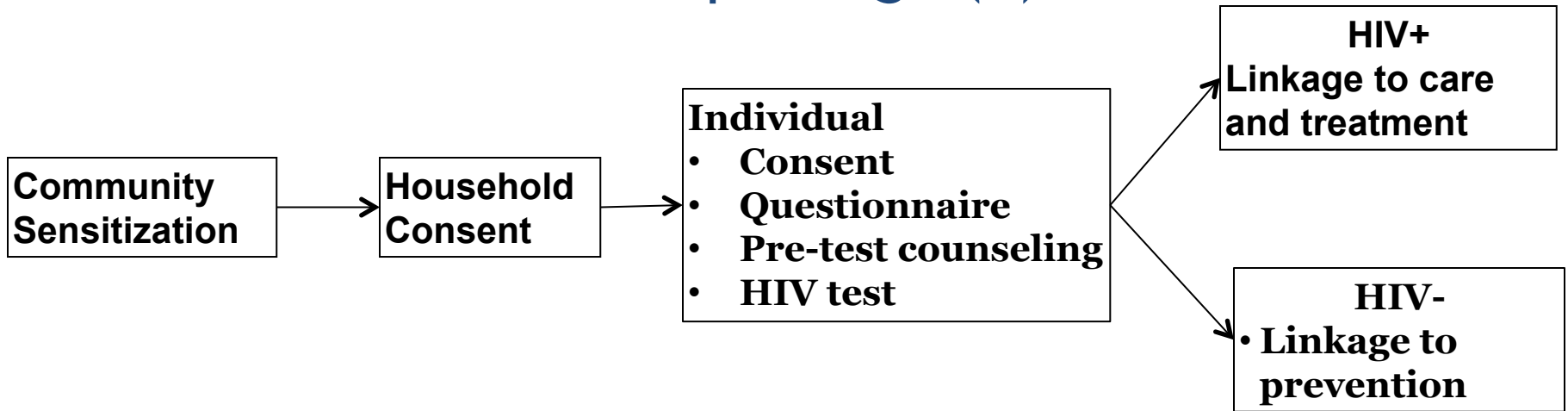
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# ART implementation



# Methods: Intervention package (1)



# Results

## Initiation of antiretroviral therapy and viral suppression after home HIV testing and counselling in KwaZulu-Natal, South Africa, and Mbarara district, Uganda: a prospective, observational intervention study

*Ruanne V Barnabas, Heidi van Rooyen, Elioda Tumwesigye, Pamela M Murnane, Jared M Baeten, Hilton Humphries, Bosco Turyamureeba, Philip Joseph, Meighan Krows, James P Hughes, Connie Celum*

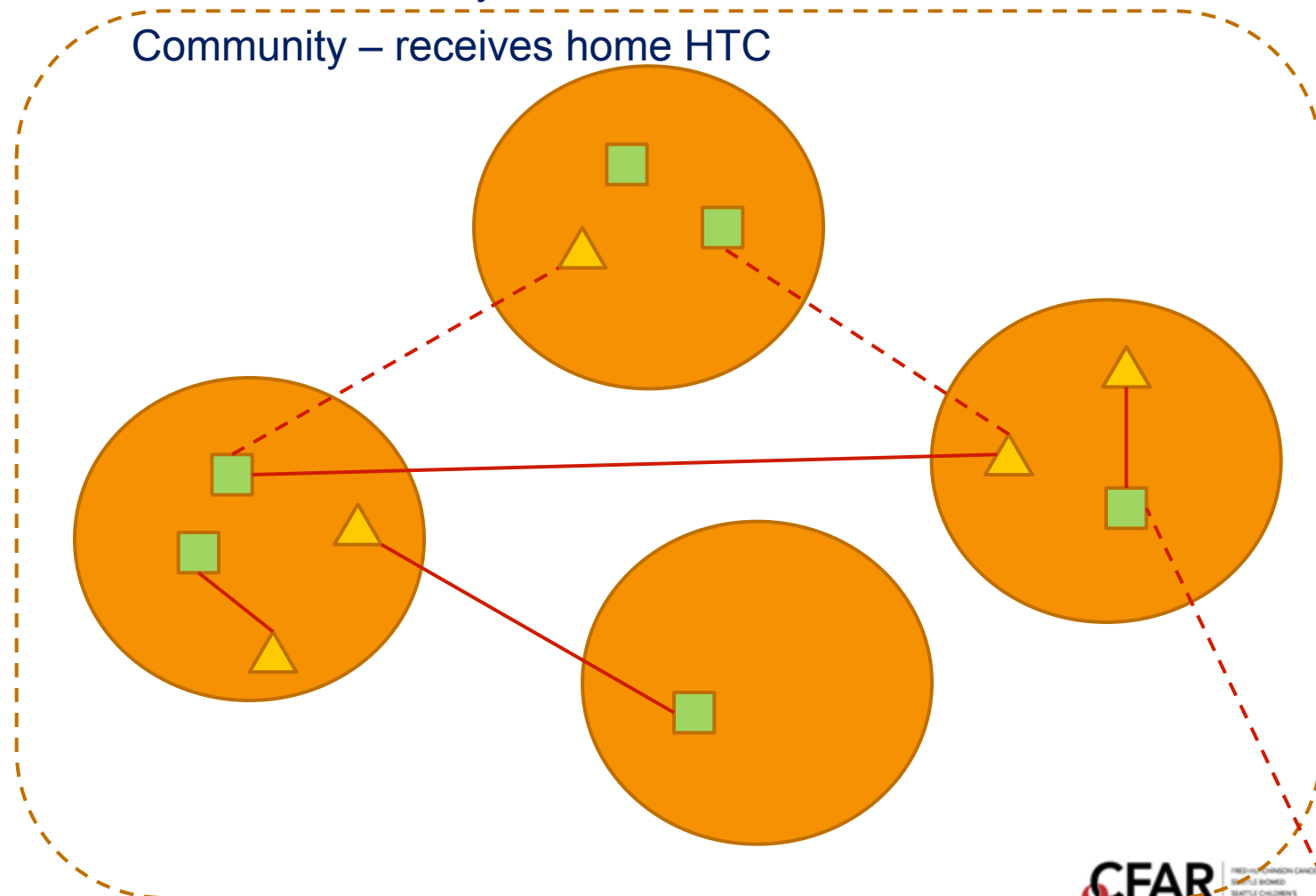
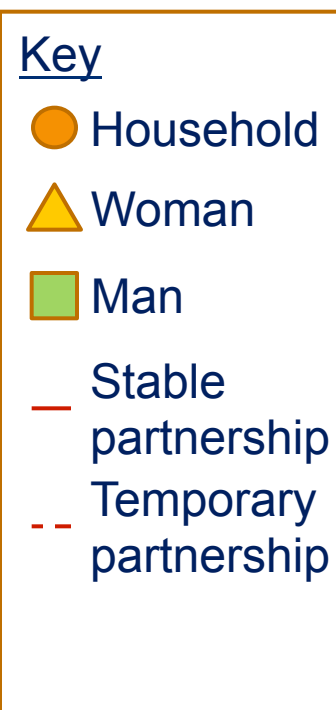
- Ankole region, southwest Uganda, and KwaZulu-Natal, South Africa
- Sept. 2011 – May 2013

Findings	N (%)
Adults tested	3,393 (96%)
HIV+ identified	635 (19%)
Visited a clinic by month 12	96%
Started ART by month 12 (among those eligible for ART)	74%
Virally suppressed by month 12 (among those on ART)	77%

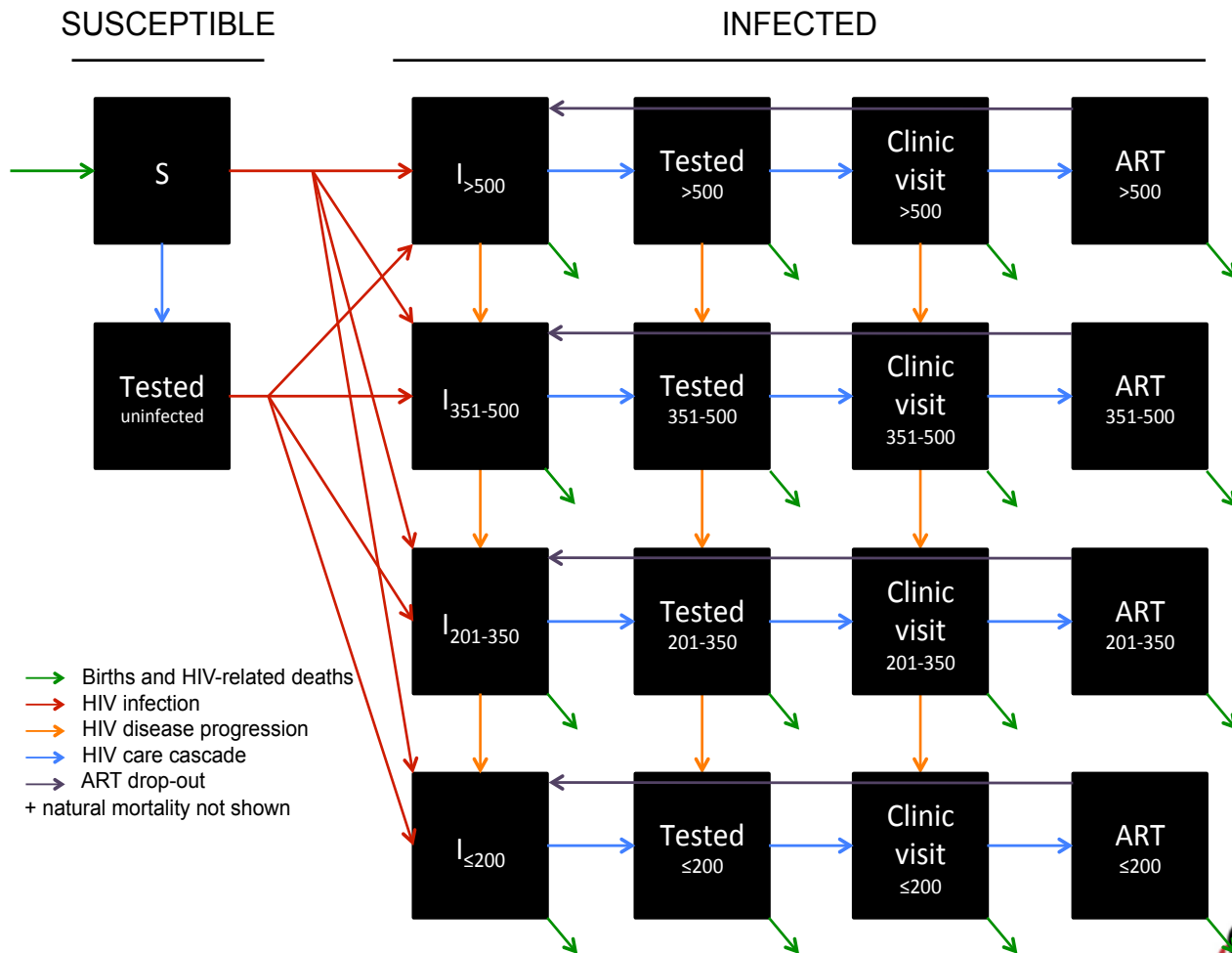
# Model: community structure & partnerships

Outside community – no intervention

Community – receives home HTC

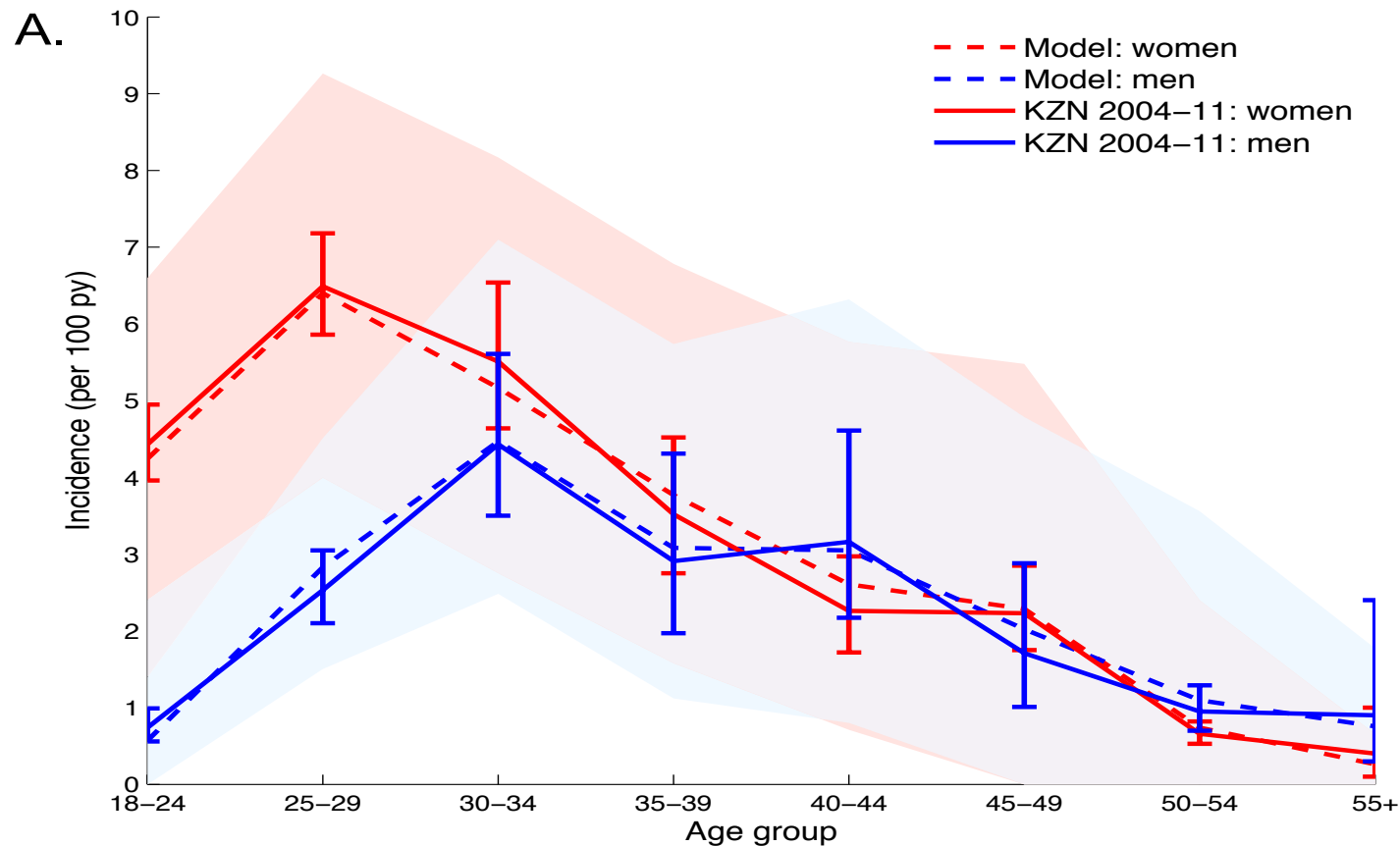


# Individual-based model structure





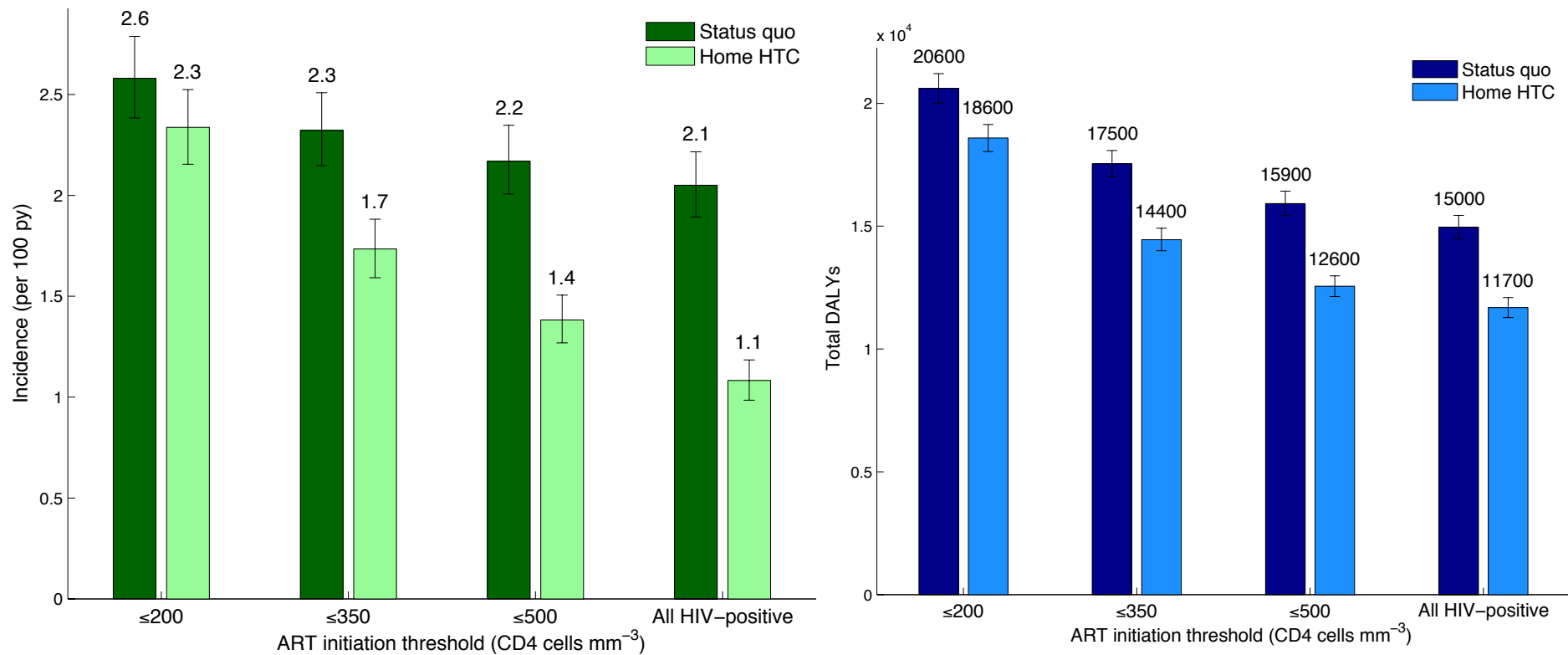
# Model prediction compares well with observed data



# Cost-effectiveness of community-based strategies to strengthen the continuum of HIV care in rural South Africa: a health economic modelling analysis

Jennifer A Smith, Monisha Sharma, Carol Levin, Jared M Baeten, Heidi van Rooyen, Connie Celum, Timothy B Hallett, Ruwan V Barnabas

## Home HTC and linkage has the potential to decrease HIV incidence



- Under new South African ART initiation criteria (CD4 ≤500 cells per  $\mu$ L), home HTC and linkage has the potential to reduce HIV incidence by 36% and total DALYs by 21% over 10 years.

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# What data do we need for models?

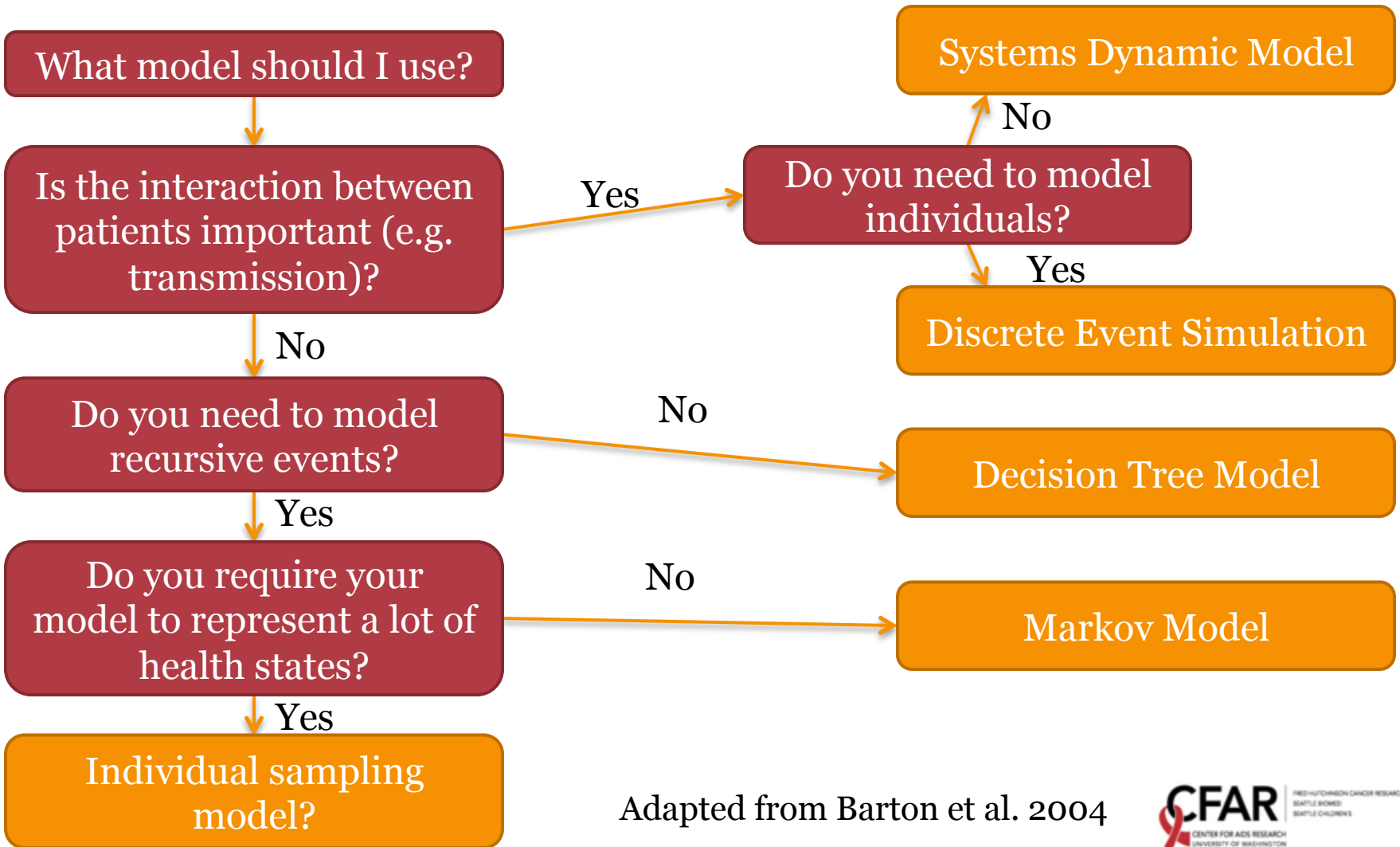
- Demographics
- Mixing patterns
- Natural history
- Transmission probability
- Factors that change susceptibility
- Factors that change infectiousness
- Effectiveness of interventions
- Engagement in health care



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# How to choose the appropriate model for health outcomes



Adapted from Barton et al. 2004

# Summary

- Infectious disease modeling is a useful tool – assumptions are explicit, characterize uncertainty
- Study data can be used to parameterize models
- Models can be used to estimate health outcomes
- Consult with a health economist and/or modeler to choose an appropriate model to answer your question
  
- Contact: [rbarnaba@uw.edu](mailto:rbarnaba@uw.edu)

# Thank you

## Study Participants ICOBI and HSRC Staff

Connie Celum, Carol Levin, Jared Baeten, Roger Ying, Aditya Khanna, Monisha Sharma, Sarah Roberts, Susie Cassels, Jim Hughes, Geoff Garnett, Meighan Krows, Hilton Humphries, Bosco Turyamureeba, Katherine Murray, Elioda Tumwesigye, Heidi van Rooyen & Judy Wasserheit



### Funding

NIH NCRR Grant 5 KL2 RR025015  
NIH CFAR Grant P30 AI027757



UNIVERSITY OF WASHINGTON  
INTERNATIONAL CLINICAL RESEARCH CENTER  
PreventionRX Study