

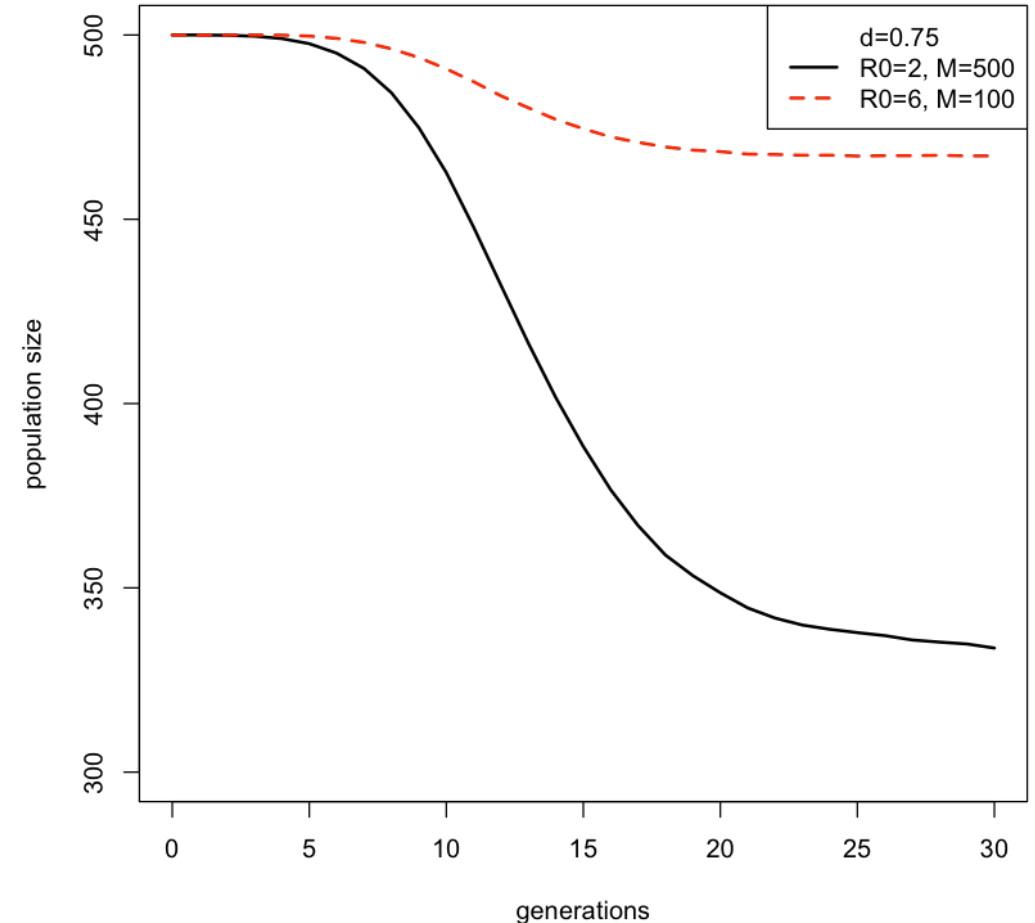
Estimating *Anopheles gambiae* R_m using insecticide and bed net field trial data

Using ecological modelling and MLE
inference to inform gene drive models.

Katie Willis

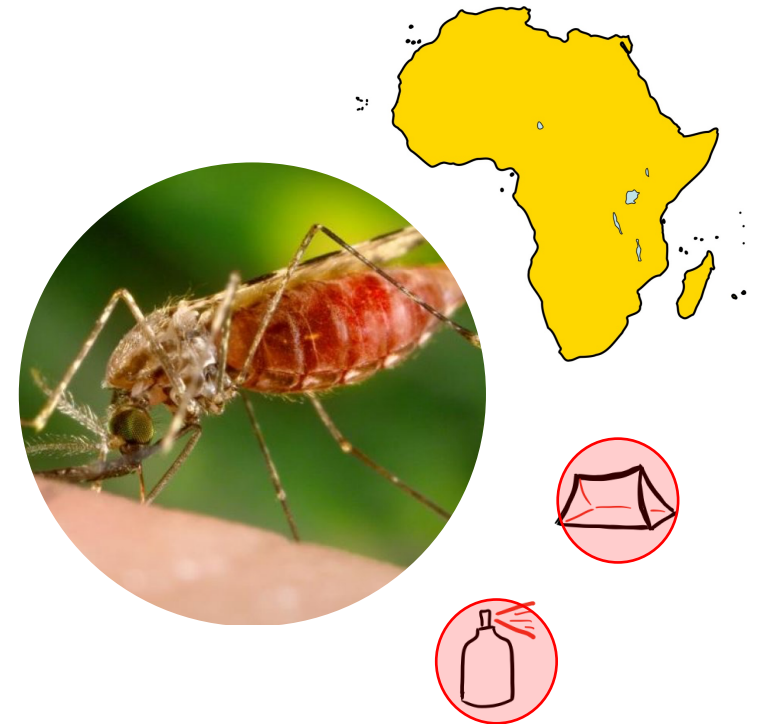
What is the R_m of wild populations of mosquitoes?

- We have seen that populations with different R_0 (R_m) respond differently to the same gene drive intervention.
 - To design a HEG/transgene which reduces the population by a desired amount we first need to know the R_m of the target population.



What is the R_m of wild populations of mosquitoes?

- We have seen that populations with different R_0 (R_m) respond differently to the same gene drive intervention.
 - To design a HEG which reduces the population by a desired amount we first need to know the R_m of the target population.
- Can we estimate R_m using data from existing field trials evaluating the impact of other interventions on density?
 - Use **Ecological modelling** and **maximum likelihood estimation** to estimate R_m and inform gene drive models.



- ① How intervention trial data contains information on R_m
- ② Model of An gambiae population dynamics
- ③ Intervention response field data
- ④ Maximum likelihood estimation

- ① How intervention data contains information on R_m ?
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What is Rm?

Beverton-holt model

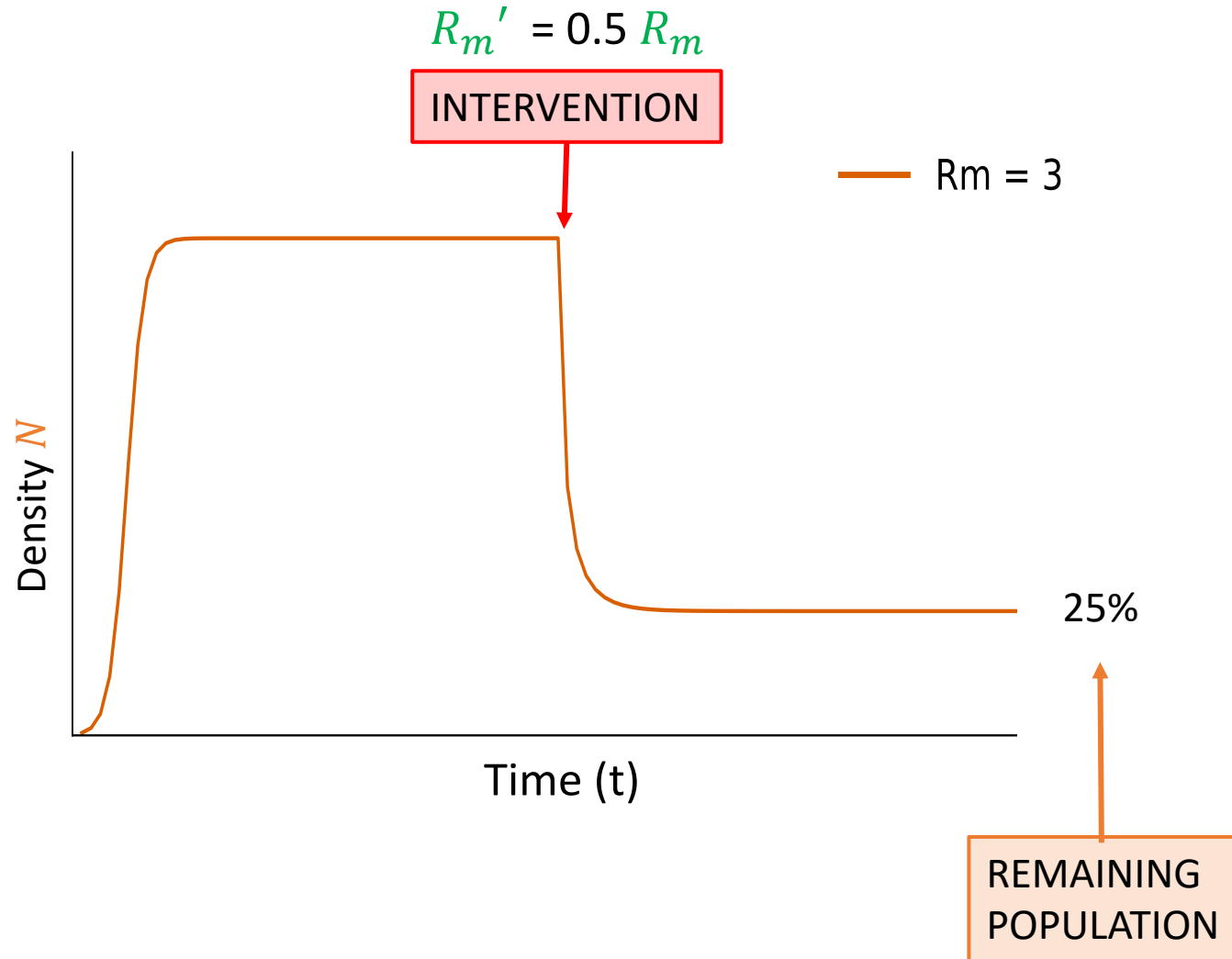
$$N_{t+1} = \frac{N_t R_0}{1 + \frac{N_t}{m}} = N_t R_0 \frac{m}{m + N_t}$$

$$N_{t+1} = N_t R_m \frac{\alpha}{\alpha + N_t}$$

R_m = Intrinsic
growth rate

Density dependent
regulation

Rm estimation using intervention data



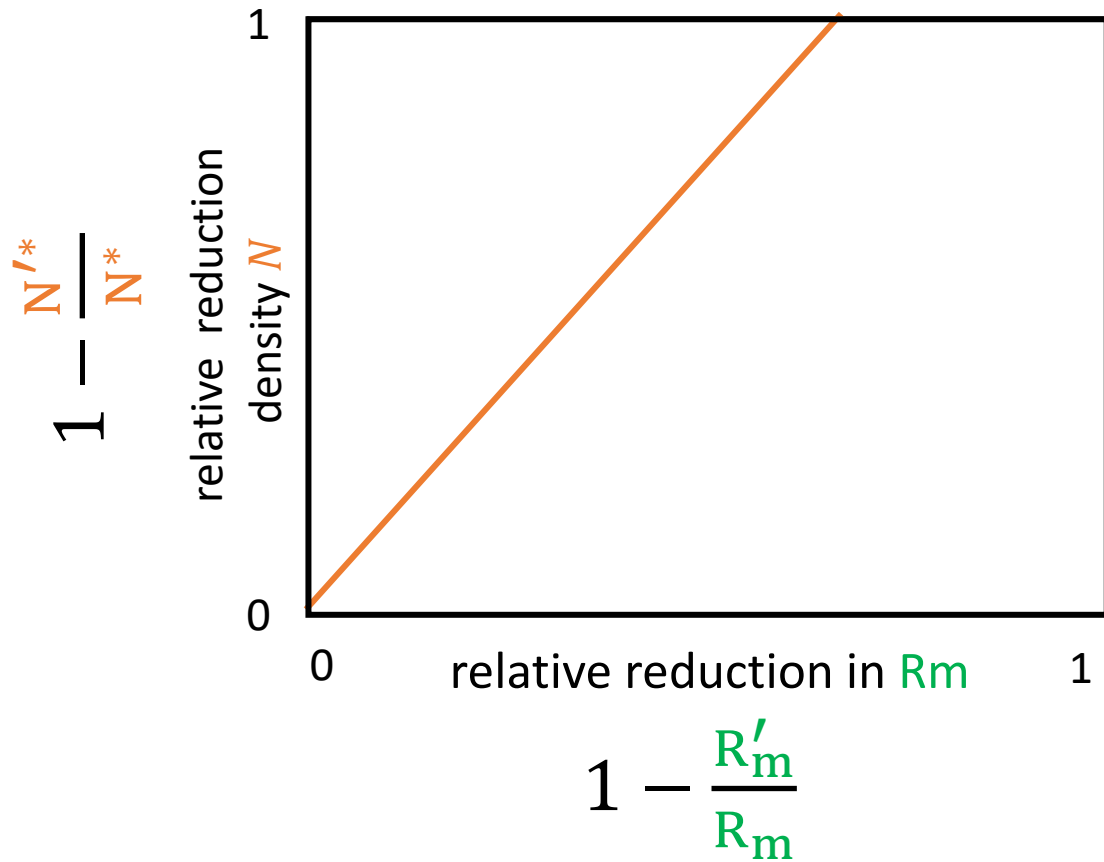
Before intervention

$$N_{t+1} = N_t R_m \frac{\alpha}{\alpha + N_t}$$

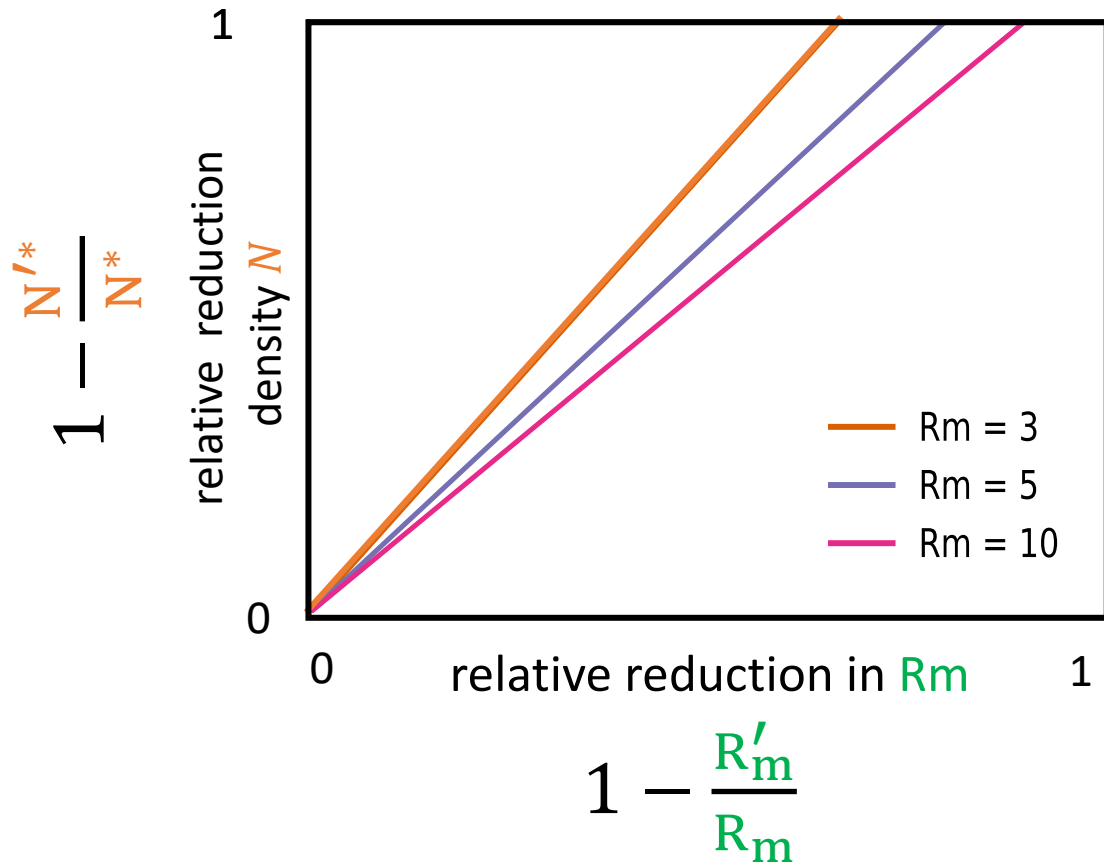
After intervention

$$N_{t+1} = N_t R_m' \frac{\alpha}{\alpha + N_t}$$

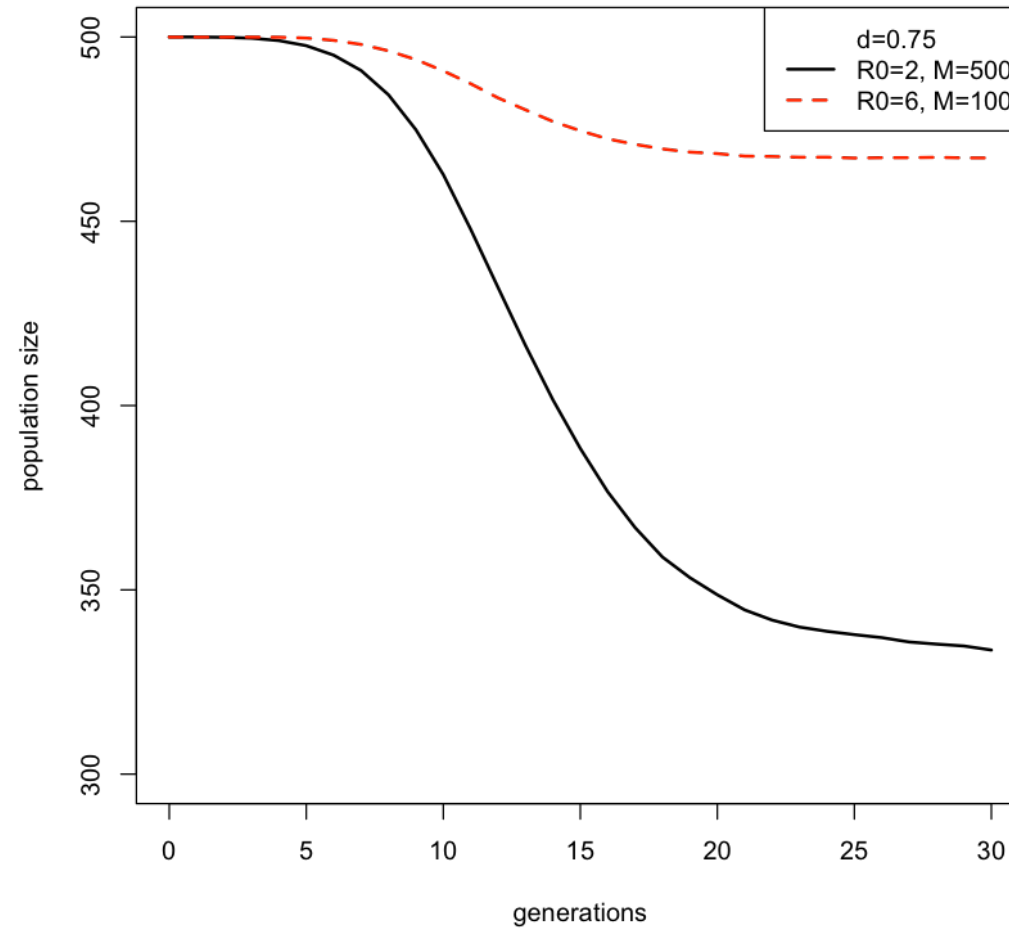
R_m estimation using intervention data



Rm estimation using intervention data

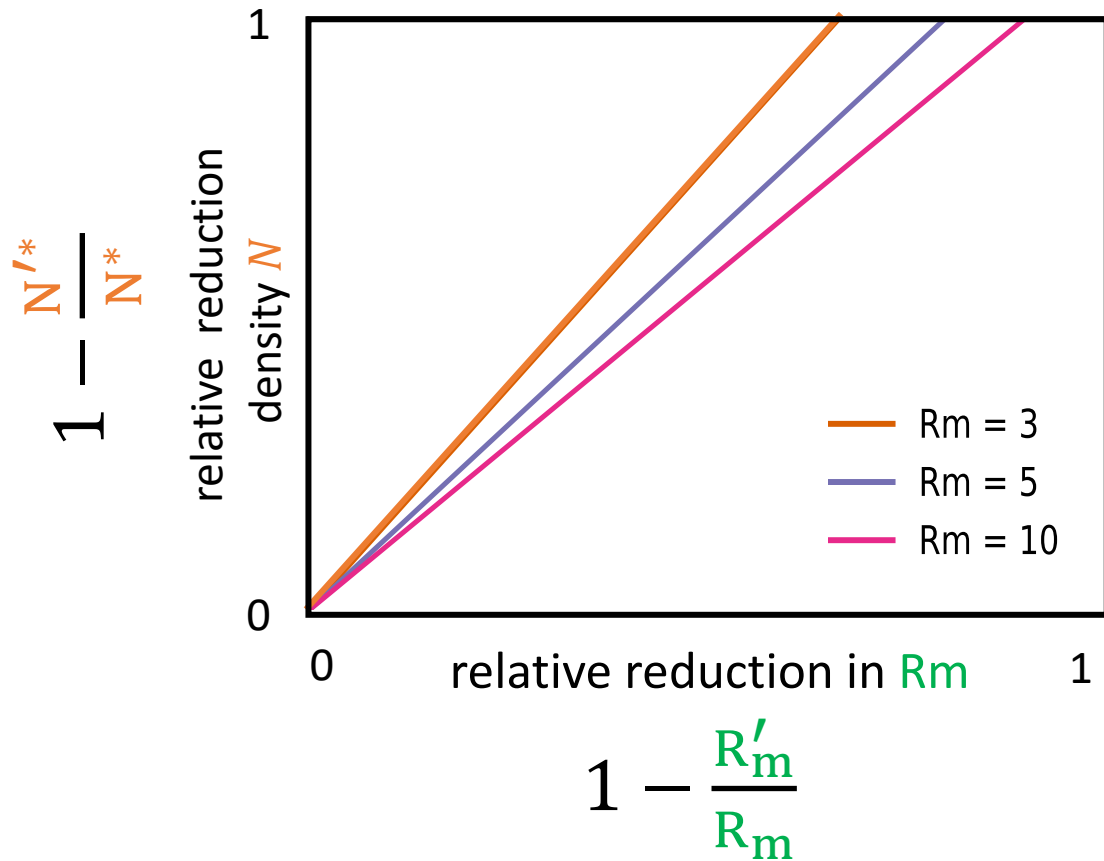


For a given reduction in intrinsic growth rate (R_m), a population with lower R_m will show a greater reduction in density (N).



This is what we saw using our gene drive simulations!

R_m estimation using intervention data



For a given reduction in intrinsic growth rate (R_m), a population with lower R_m will show a greater reduction in density (N).

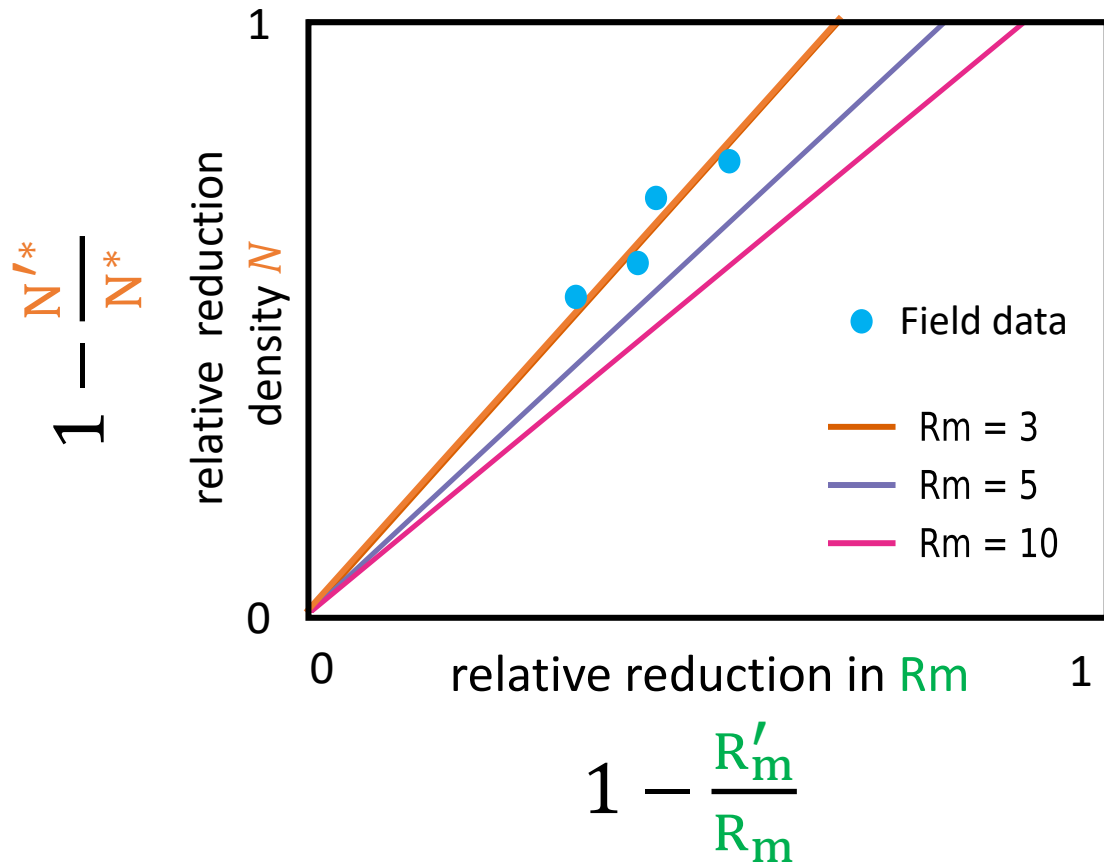
- **A model of population dynamics**

Relating a change in R_m to a change in N at equilibrium

$$N^* = (R_m - 1)\alpha$$

$$N'^* = (R'_m - 1)\alpha$$

Rm estimation using intervention data



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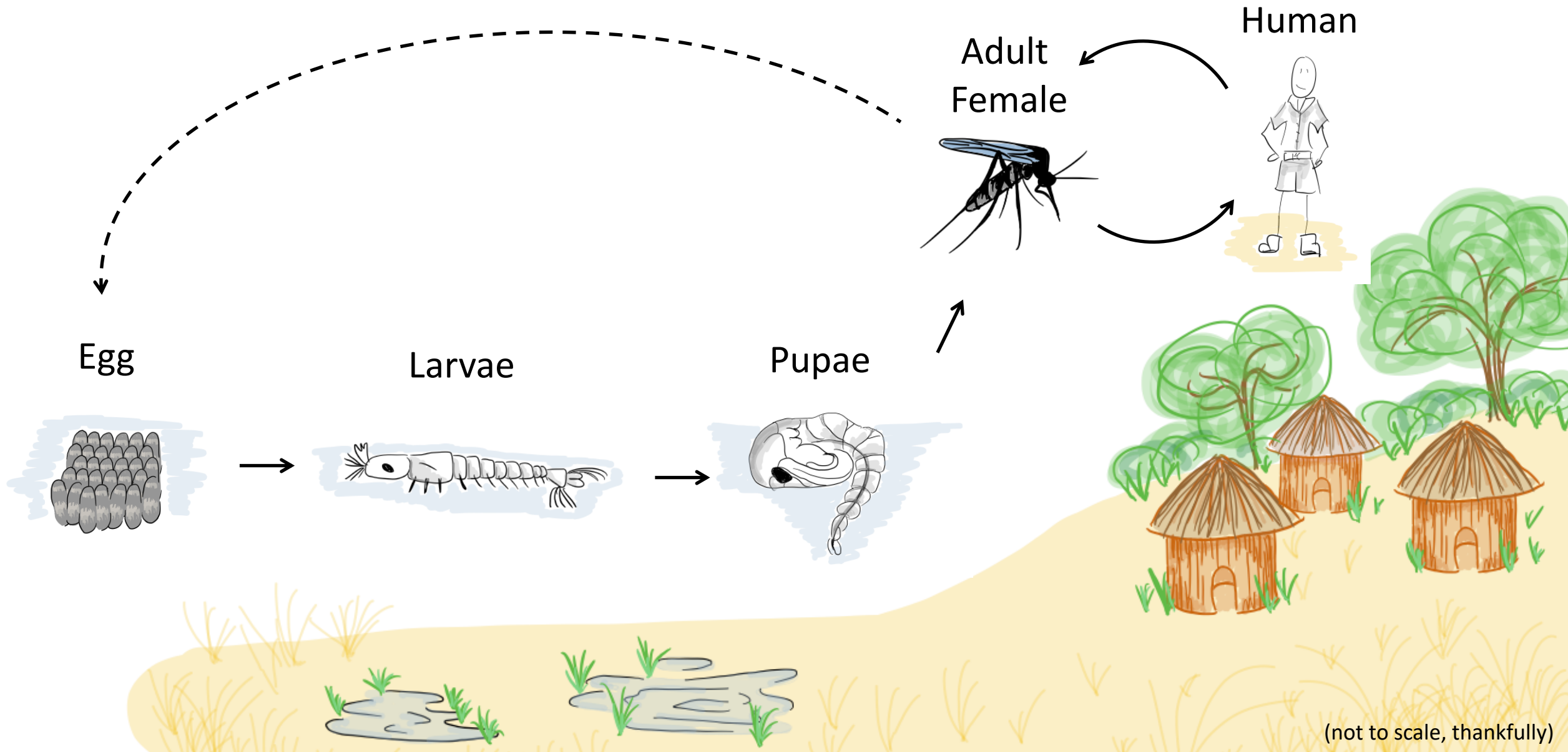
- **Intervention field data**

x Effect on the intrinsic growth rate R_m

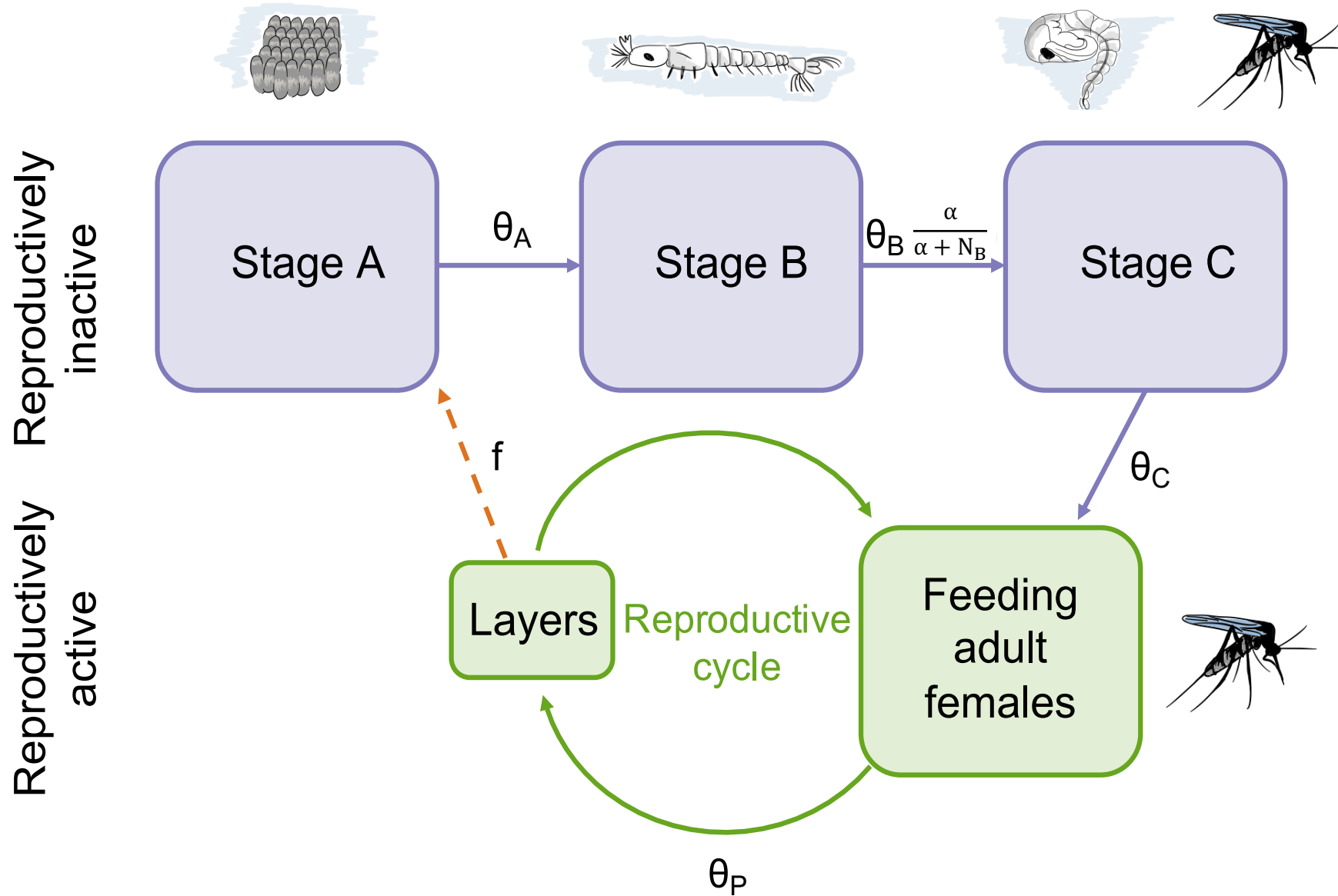
y Change in equilibrium population density N

- ① How intervention data contains information on R_m ?
- ② Model of An gambiae population dynamics
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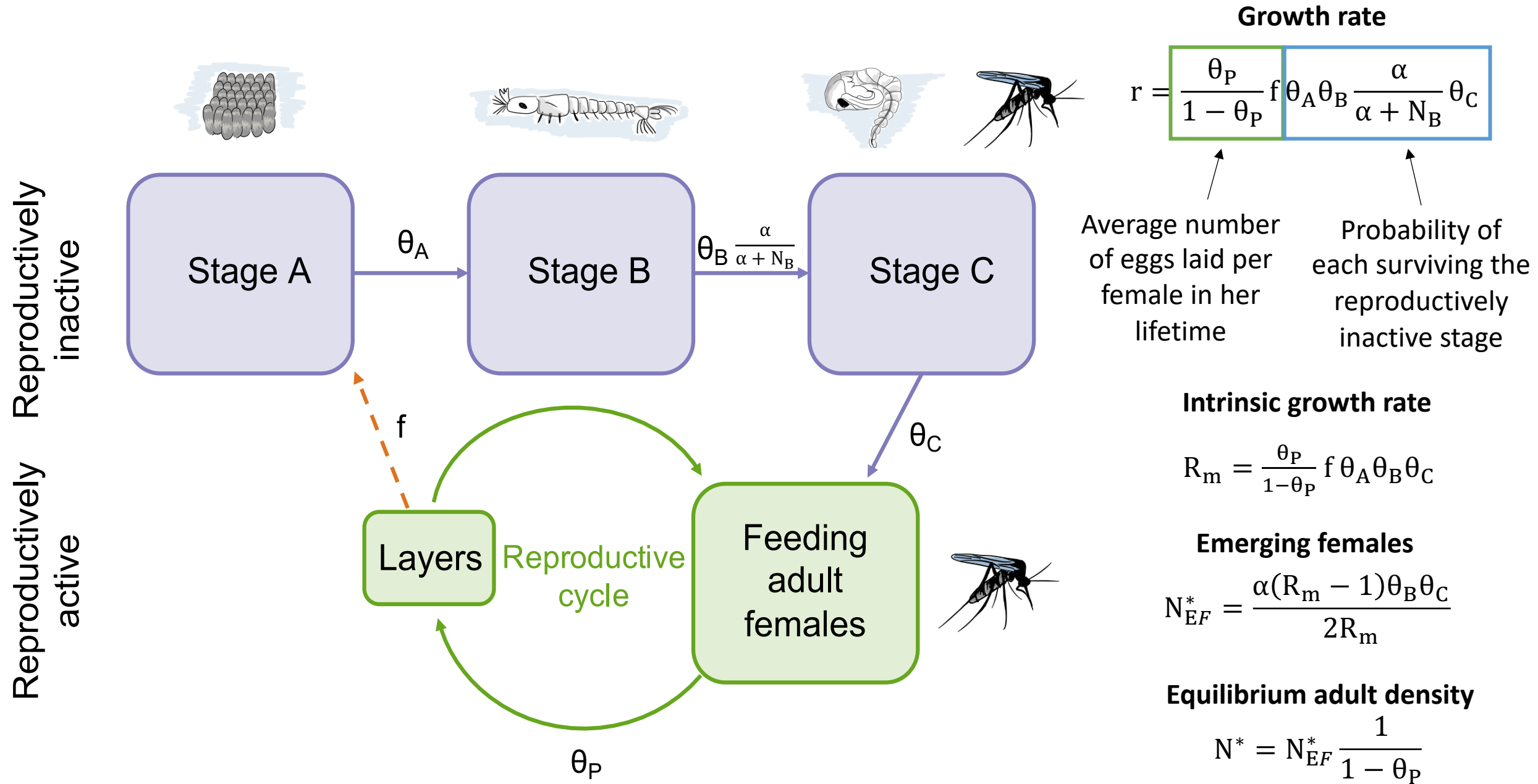
Using a model of An gambiae population dynamics



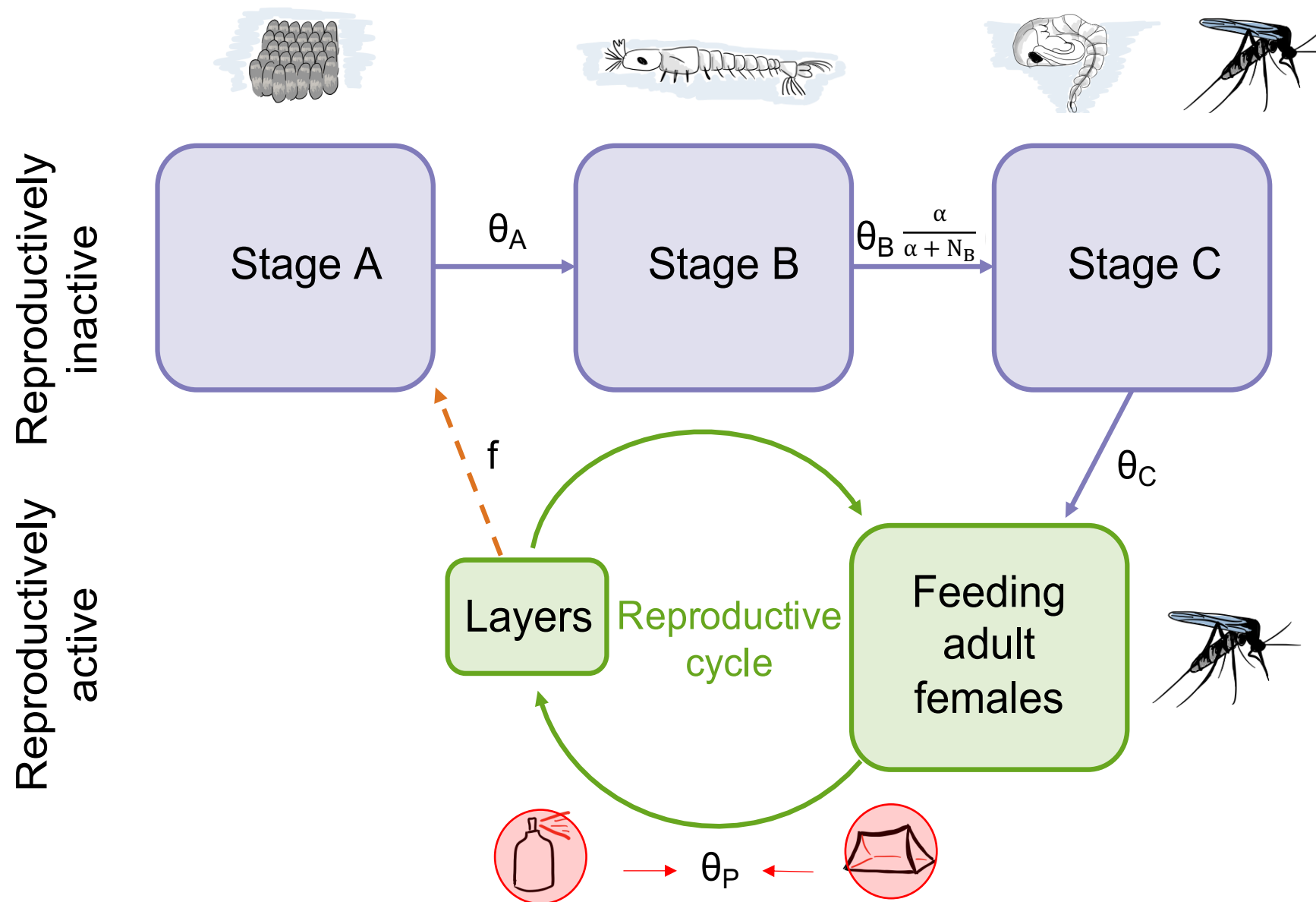
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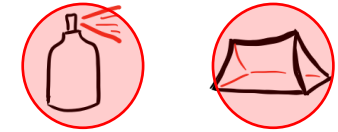


Using a model of An gambiae population dynamics



We are interested in interventions which modify R_m

$$R_m = \frac{\theta_P}{1 - \theta_P} f \theta_A \theta_B \theta_C$$



Insecticide bed nets and sprays act to reduce R_m by reducing adult survival (θ_P)

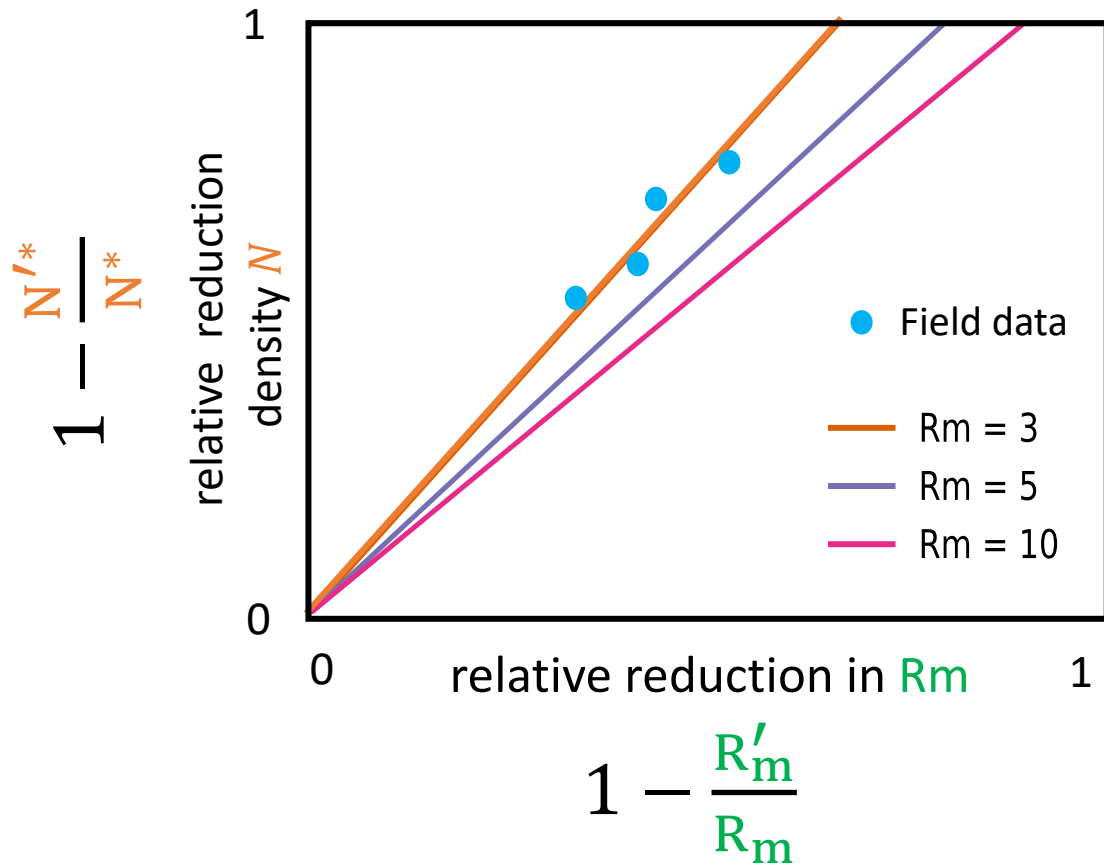
This changes the average number of egg laying cycles by

$$\frac{\theta_P}{1 - \theta_P} \text{ to } \frac{\theta_P'}{1 - \theta_P'}$$

And therefore reduces R_m by

$$1 - \frac{R_m'}{R_m} = 1 - \frac{(1 - \theta_P)\theta_P'}{(1 - \theta_P')\theta_P}$$

R_m estimation using intervention data



For a given reduction in intrinsic growth rate (R_m), a population with lower R_m will show a greater reduction in density (N).



A model of population dynamics

Relating a change in R_m to a change in N at equilibrium

- Intervention field data

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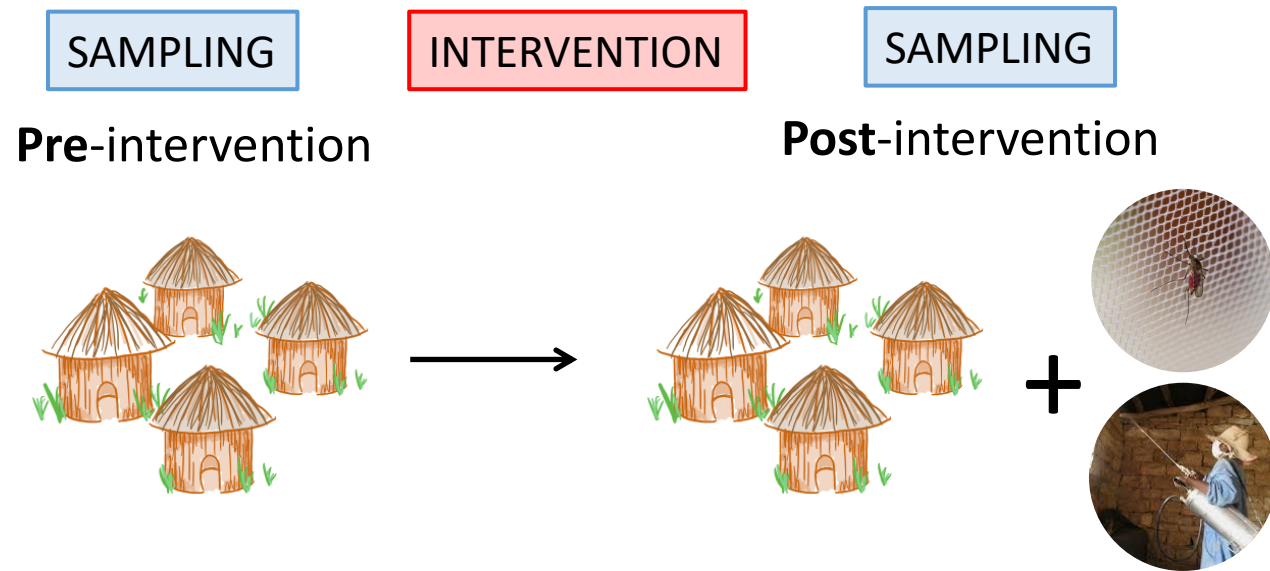
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Intervention response field data

\mathcal{X} Effect on the intrinsic growth rate R_m ($\theta_p, \theta_{p'}$)

\mathcal{Y} Change in equilibrium population density N



Intervention response field data

\mathcal{X} Effect on the intrinsic growth rate R_m ($\theta_p, \theta_{p'}$)

$$1 - \frac{R'_m}{R_m} = 1 - \frac{(1 - \theta_p)\theta_{p'}}{(1 - \theta_{p'})\theta_p}$$

\mathcal{Y} Change in equilibrium population density N

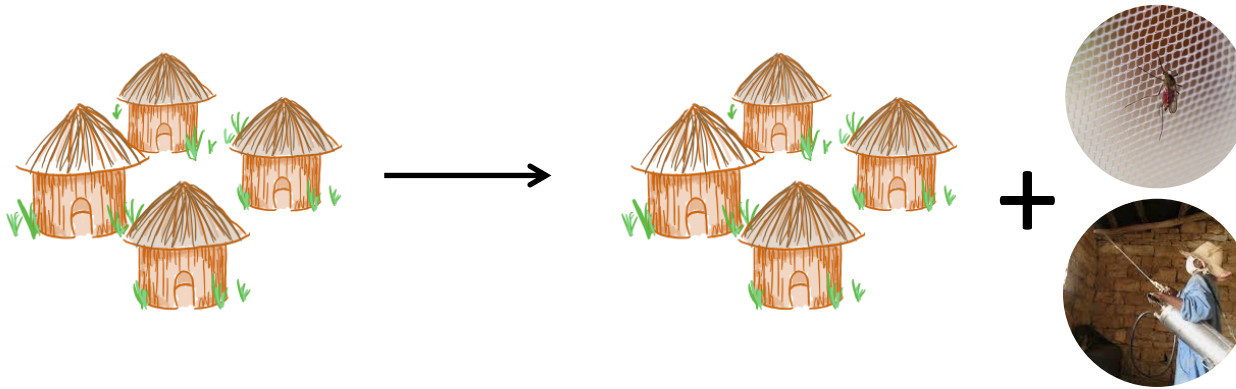
SAMPLING

INTERVENTION

SAMPLING

Pre-intervention

Post-intervention



P = number tested for parity

ρ = number parous

$\theta_p = \frac{\rho}{P}$ = parous rate pre-intervention

P' = number tested for parity

ρ' = number parous

$\theta_{p'} = \frac{\rho'}{P'}$ = parous rate post-intervention

Intervention response field data

\mathcal{X} Effect on the intrinsic growth rate R_m ($\theta_p, \theta_{p'}$)

$$1 - \frac{R'_m}{R_m} = 1 - \frac{(1 - \theta_p)\theta_{p'}}{(1 - \theta_{p'})\theta_p}$$

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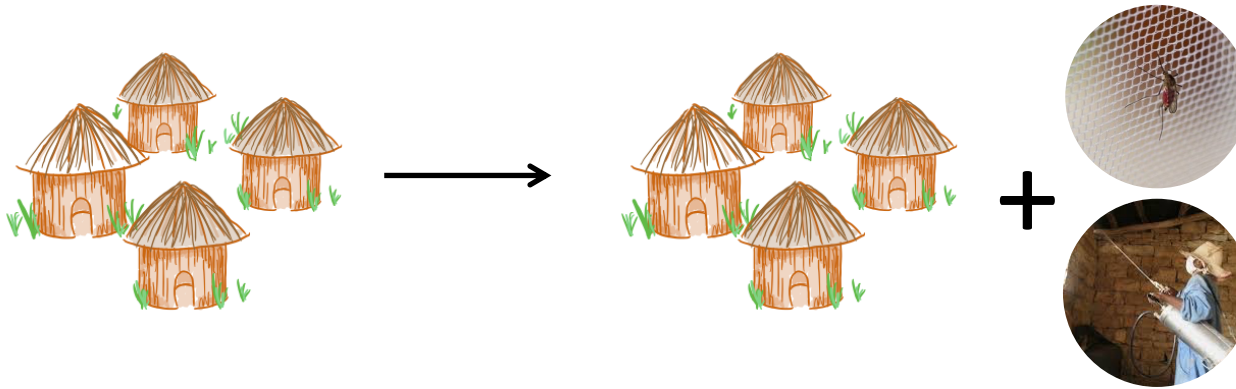
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n = number of mosquitoes
sampled pre-intervention

n' = number of mosquitoes
sampled pre-intervention

Intervention response field data

\mathcal{X} Effect on the intrinsic growth rate R_m ($\theta_p, \theta_{p'}$)

$$1 - \frac{R'_m}{R_m} = 1 - \frac{(1 - \theta_p)\theta_{p'}}{(1 - \theta_{p'})\theta_p}$$

\mathcal{Y} Change in equilibrium population density N

$$1 - \frac{N'^*}{N^*} = 1 - \frac{n'}{n}$$

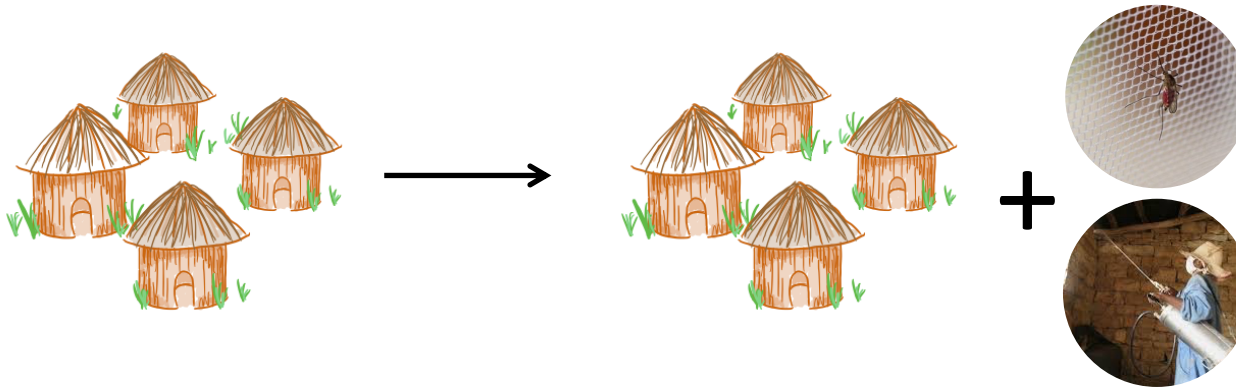
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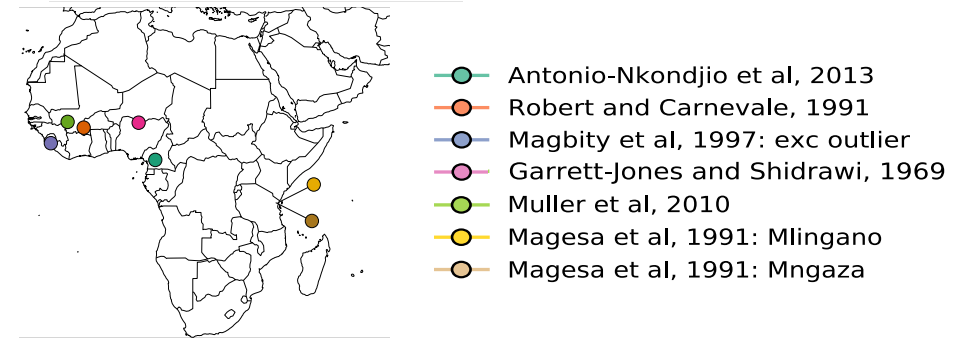
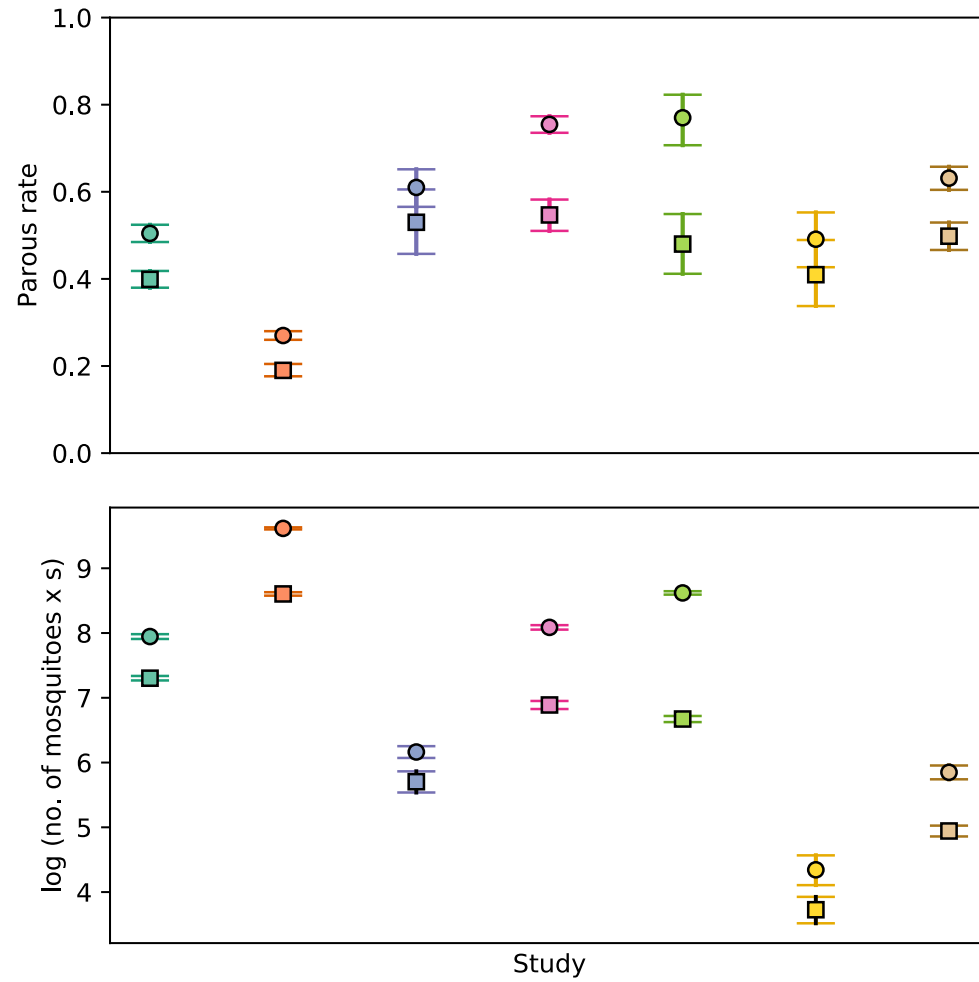
P' = number tested for parity

ρ' = number parous

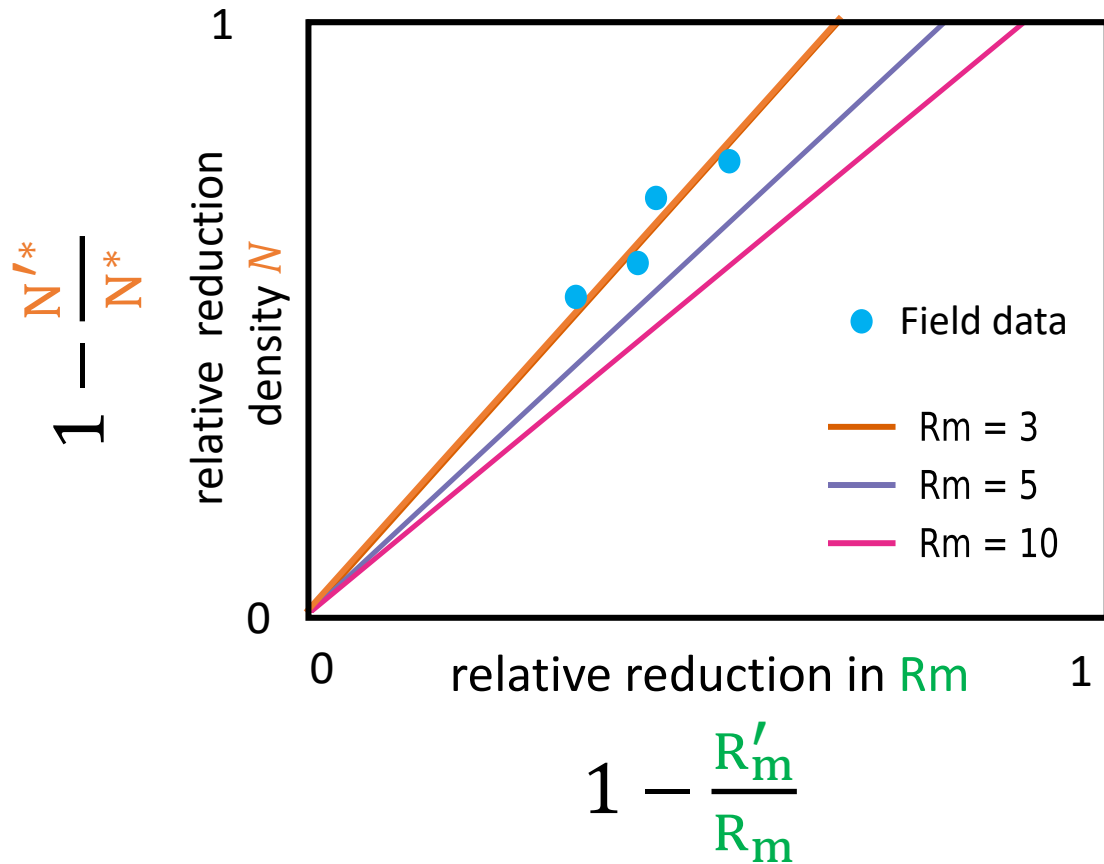
$\theta'_p = \frac{\rho'}{P'}$ = parous rate post-intervention

n = number of mosquitoes
sampled pre-intervention

n' = number of mosquitoes
sampled pre-intervention



Rm estimation using intervention data



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A model of population dynamics

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Intervention field data

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Maximum likelihood estimation: REMINDER

MLE

- Maximum Likelihood estimation (MLE) is a method to estimate parameters of a statistical model
- When the method is applied to a **dataset** with a statistical **model**, MLE provides estimates for the associated **parameters**.
- “The parameter values that makes the observed dataset most *probable*.”

$$L(\theta|x) = p(x|\theta)$$

- The likelihood function $L(\theta|x)$ is used to quantify how likely the parameter values (θ) are given some observation (x).
- Treat the parameters as unknown
- For each sample (x) and a given model, let θ be a parameter value at which $L(\theta|x)$ attains a maximum.
- This is then the maximum likelihood estimate of θ for x

Maximum likelihood estimation:

Our statistical model:

To build our likelihood function we first need to make some assumptions about the sampling probability of our data.

Parous rates

P = number tested for parity

ρ = number parous

$\theta_p = \frac{\rho}{P}$ = parous rate pre-intervention

Binomial distribution

$$\theta_P \sim \text{Bin}(\rho, P)$$

Sum of P independent and identically distributed Bernoulli random variables, where a Bernoulli r.v. is a binary outcome (success or fail)

Maximum likelihood estimation:

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Mosquito counts

n = number of mosquitoes sampled pre-intervention

Ns = total number of feeding mosquitoes * sampling probability

Poisson distribution

$$Ns \sim \text{Pois}(n)$$

Number of events occurring in a fixed interval of time.

Maximum likelihood estimation

Our statistical model: $\theta_p \sim \text{Bin}(\rho, P)$ $Ns \sim \text{Pois}(n)$

$$\log L(\text{Parameters} | \text{Data})$$

$$\begin{aligned} &= \rho \log \theta_p + (P - \rho) \log(1 - \theta_p) \longrightarrow \text{Parous rates} \longrightarrow \text{Binomial distribution} \\ &+ \rho' \log \theta_p' + (P' - \rho') \log(1 - \theta_p') \longrightarrow \\ &+ (-Ns + n \log Ns - \log(n!)) \longrightarrow \text{Mosquito counts} \longrightarrow \text{Poisson distribution} \\ &+ (-N's + n' \log N's - \log(n'!)) \longrightarrow \end{aligned}$$

Incorporate the model of mosquito population dynamics:

$$N = \frac{\alpha(R_m - 1)\theta_B\theta_C}{2R_m(1 - \theta_P)} \quad N' = \frac{\alpha(R_m' - 1)\theta_B\theta_C}{2R_m'(1 - \theta_P')}$$

$$\log L(R_m, \theta_p, \theta_p', k | \rho, P, n, \rho', P', n')$$

$$k = \alpha\theta_B\theta_C s$$

Parameters

θ_p = inferred parous rate

N = equilibrium number of mosquitoes

s = sampling probability

Data

n = number of caught mosquitoes

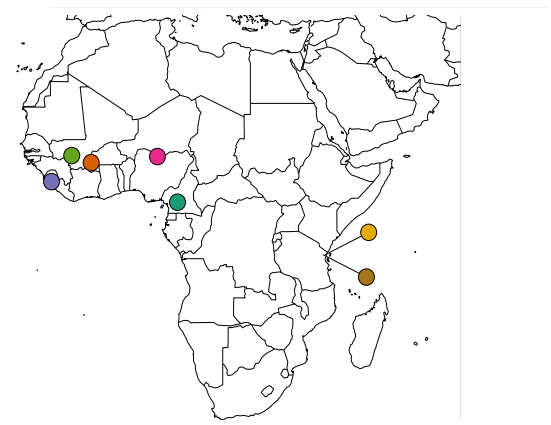
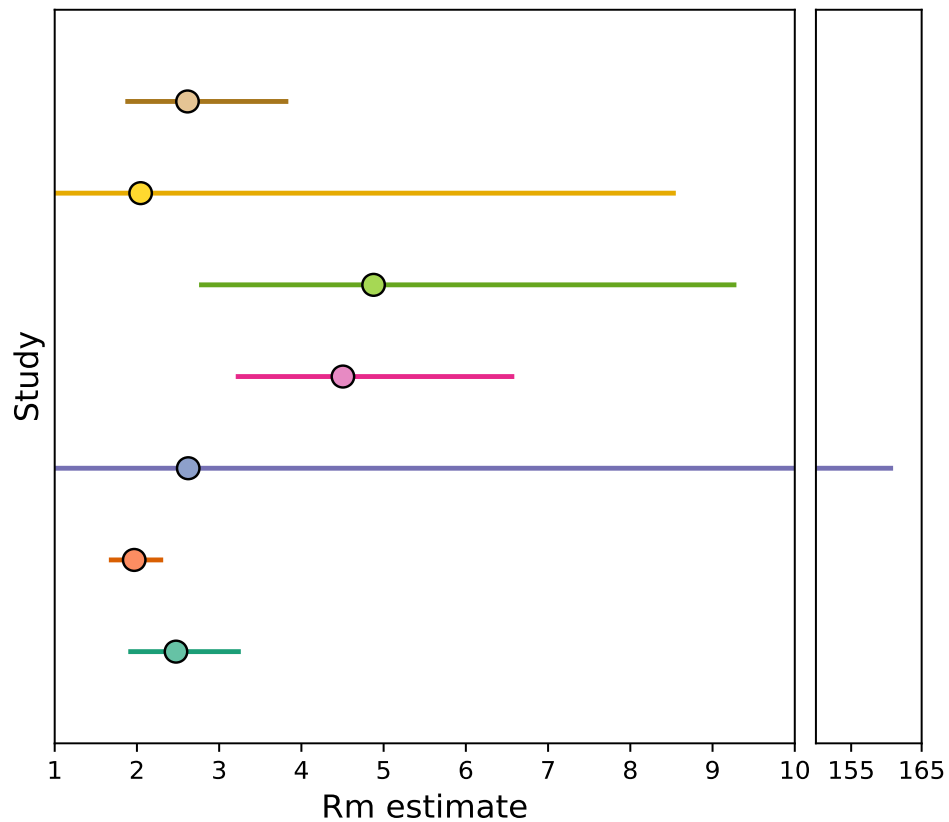
ρ = number of parous mosquitoes

P = number of dissected mosquitoes

' denotes post intervention

Maximum likelihood estimation: Maximum likelihood estimates

$$\log L(Rm, \theta_p, \theta'_p, k | \rho, P, n, \rho', P', n')$$



- Antonio-Nkondjio et al, 2013
- Robert and Carnevale, 1991
- Magbity et al, 1997: exc outlier
- Garrett-Jones and Shidrawi, 1969
- Muller et al, 2010
- Magesa et al, 1991: Mlingano
- Magesa et al, 1991: Mngaza

Conclusions

- Using an ecological model we can see that observations of changes in density in response to interventions can provide information of the intrinsic growth rate of a population
- MLE can be applied to a **dataset** using a **statistical model** to provide estimates for the associated **parameters**
- Here I apply MLE methods to **intervention field trial data** using a **model of mosquito population dynamics** to estimate the **R_m of the population**

MSc/MRes Project

Modelling the population dynamics of *Anopheles gambiae* mosquitoes.

Anopheles gambiae mosquitoes are the primary vectors of malaria and despite extensive application of population control strategies, the disease remains a substantial health burden. An improved understanding of *An. gambiae* population dynamics is required to predict, evaluate and optimize the impact of intervention strategies on disease transmission. Of particular interest is the intrinsic population growth rate, a parameter which influences the strength of intervention required to modify population densities. Little is known about *An. gambiae* growth rate in the field, and current estimations have relied on collecting detailed demographic data, often measured under laboratory conditions. The project involves modifying and extending a statistical inference method developed by our group for estimating the intrinsic growth rate using field measurements of population densities in response to control strategies.

This is a fully theoretical project and is likely to involve:

- Extension of the current population dynamics model to include additional features e.g. **seasonality** and/or **migration**.
- Use of maximum likelihood/Bayesian approaches to infer population biology parameters from the data.
- Use of simulations to assess how well the inference method is able to estimate parameters under different ecological scenarios.
- Application of the method to suitable field data identified from the literature.

The project will require basic computational/programming skills and will provide opportunity to develop skills in mathematical modelling and maximum likelihood/Bayesian methods for parameter inference.