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Reasoning

Reasoning Infectious

Intervention

The Role of Mathematical Models in Vaccine Evaluation ENSP July 2013

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# **Epidemiological Reasoning**



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Figure: Exposure definition, outcome definition, link function, sources of uncetainties

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# Realism, Generality, Precision: Legitimate vs Illegitimate Simplifications



Figure: Paradigm "tension"

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# Useful quotes

Far better an approximate answer to the right question, which is often vague, than an exact answer to the wrong question, which can always be made precise (John W. Tukey).

No matter how beautiful your theory, no matter how clever you are or what your name is, if it disagrees with experiment, it's wrong (Richard Feynman).

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Entities should not be multiplied beyond necessity (Occam's Razor)

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### Reasoning

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Basics

Example Epidemiological Categorie

Incidence Rates

Recovery Rate

Immunity Rate

Demographic Rates

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# Parameters of interest



Figure: Basic parameters (Grassly 2008)

Image: Image:

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Reasoning Infectious Basics Example Epidemiological Cat

Incidence Hates

Recovery Rates

Immunity Hates

Demographic Rates

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# $R_0$ as a threshold parameter

 $R_0 < 1$  $p_c = \frac{R_0 - 1}{R_0} = 1 - \frac{1}{R_0}$ 

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# Reasoning Infectious Basics

Example Epidemiological Categories Incidence Rates Recovery Rates Immunity Rates Demographic Rates

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# Epidemiological Reasoning



Figure: Estimating *R*<sub>0</sub> (Grassly 2008)

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Reasoning Infectious Basics Example

> Epidemiological Categories Incidence Rates Recovery Rates Immunity Rates Demographic Rates

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# Malaria Life Cycle



Figure: Example: biology of malaria



Figure: Morbidity, immunity, and degree of infectiousness (super-spreader), distribution clusters

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Reasoning Infectious Basics Example Epidemiological Categories Incidence Rates Recovery Rates Immunity Rates Demographic Rates

# Epidemiological Translation of the life cycle: incidence rates



Figure: Dependence on the number of previously infected, vector density, vector competence, superinfection, within host competition, bottlenecks, incomplete immunity.

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# Epidemiological Translation of the life cycle: recovery rates



Figure: Dependence on parasite load, immune memory, superinfection, within host competition, bottlenecks.



Figure: Phenomena in various levels, time delays, immune boosting.

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# Epidemiological Translation of the life cycle: demographic rates



Figure: Migration, seasonality, age structure.

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- Classon
- Biological efficac
- Challenge
- Study Desi
- Behavior
- Hierarchy
- Causal inference

# Mechanisms of action and intervention delivery



Figure: Infection blocking, disease modifying, and transmission blocking; pulse, continuous, campaign a state of the state

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Equations

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# **Model Equations**

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$\frac{\partial x_1}{\partial t}$	=	$\delta + R_1(r_1, z_1, z_2)y_2 - (\lambda_1 + \delta)x_1 + \Delta_3(t)$
$\frac{\partial x_3}{\partial t}$	=	$R_2(r_2, z_3)y_3 - (\lambda_2 + \delta)x_3 - \Delta_3(t)$
$\frac{\partial y_2}{\partial t}$	=	$\lambda_1 x_1 - [A_1(\alpha_2, z_1, z_2) + R_1(r_1, z_1, z_2) + \delta] y_2$
$\frac{\partial y_3}{\partial t}$	=	$\lambda_2 x_3 + A_1(\alpha_2, z_1, z_2) y_2 - [R_2(r_2, z_3) + \delta] y_3$
<i>y</i> <sub>1</sub>	=	$\frac{1 - e^{-z_1}}{1 - e^{-(z_1 + z_2)}} y_2, \ \lambda_i = m_3(t - N_1)ab_i$
$\frac{\partial z_1}{\partial t}$	=	$\lambda_1 - \alpha_1 z_1,  \frac{\partial z_2}{\partial t} = \alpha_1 z_1 - r_1 z_2,  \frac{\partial z_3}{\partial t} = \lambda_2 - r_2 z_3$
<i>R</i> <sub>1</sub>	=	$r_1 z_2 \frac{e^{-(z_1+z_2)}}{1-e^{-(z_1+z_2)}}, \ R_2 = r_2 z_3 \frac{e^{-z_3}}{1-e^{-z_3}}$
$A_1(\alpha_2,z_1,z_2)$	=	$\alpha_2\left(1-e^{-(z_1+z_2)}\right)$
$\Delta_3(t)$	=	$\begin{cases} R_2(r_2, z_3, t - \tau) y_3(t - \tau) + (f - b_2) h_{\theta}(t - \tau) \end{cases}$
		$\left[x_{3}(t-\tau)-b_{2}\int_{t-\tau-N_{1}}^{t-\tau}h_{\theta}(u)x_{3}(u)du\right]\right\}e^{-(\tilde{h}_{b}+\delta)\tau}$
$\bar{h}_b$	=	$\int_{t-\tau}^t \frac{h_b(u)}{\tau} du, \ h_b(t) = m_3(t)af, \ b_2 \le t \le 1$

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Mechanisn

Equations

Glossary

**Biological efficacy** 

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Behavior

Hierarchy

Causal inference

# Glossary

Α. rate at which immunity to P. falciparum infection is acquired by a human host prop. of bites by infec. mosquitoes on neg. nonimmune hosts actually resulting in infec. b, prop. of bites by infectious mosquitoes on neg. immune hosts actually resulting in infec. h boosting factor (prop. of bites by infec, mosg, on immune hosts that boost immunity) entomologic inoc. rate (EIR): # of P. falciparum infectious bites per human per day N. P. falciparum parasitemia incubation period in humans (in days) rate ctt of elimination of a brood of parasites by nonimmune positive hosts (in days -1) r. rate constant of elimination of a brood of parasites by immune positive hosts (in days - 1) recovery rate for nonimmune positive individuals (in days -1) R. recovery rate for immune positive individuals (in days -1) R  $X_1$ proportion of nonimmune negative (i.e., naive) individuals in the population Χ. proportion of immune negative individuals Υ. proportion of nonimmune positive individuals potentially infectious for mosquitoes proportion of nonimmune positive individuals Y<sub>2</sub> proportion of immune positive individuals average number of infectious broods of the parasite per nonimmune positive human host Ζ. average number of noninfectious broods of the parasite per nonimmune positive human host average number of noninfectious broods of the parasite per immune positive human host recovery rate from infectiousness to mosquitoes among nonimmune positive hosts (in days - 1)  $\alpha$ . maximum rate at which immunity to P. falciparum infection can be acquired (in days - 1)  $\alpha_{2}$ death and birth rate in the human population (in days -1) δ  $\Delta_3$ daily fraction of immune negative subjects losing immunity  $\lambda_{1}$ infection rate for nonimmune negative subjects (probability per day of such a subject's becoming infected) λ. infection rate for immune negative subjects (probability per day of such a subject's becoming infected)

au time delay needed for an immune host to lose immunity in the absence of exposure to infection (in days)

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Equations

Glossary

**Biological efficacy** 

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Study Des

Behavior

Hierarchy

Causal inference

# Measure of intervention efficacy

In the lab: # of cases # of cases V, E, PE KC, UC NV, E, PE KC UC In the field: # of cases # of cases NV, E?, PE? , V, E?, PE?

V, NV - treatment (vaccination)

E, PE - transmission level, previous exp

KC, UC - known and unknown covaritates

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- Intervention
- Mechanism
- Equations
- Glossary
- Biological efficacy
- Challenge
- Study Des
- Behavior
- Hierarchy
- Causal inference

# Challenge: Exposure to infection





DESIGN IIb direct + indirect

Figure: Direct, indirect and overall effects

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# Behavior of common measures of intervention efficacy



FIGURE 1a Incidence (effective inoculation rate) in non-immune susceptibles of populations A and B during the first two years of observation. FIGURE 1b Incidence density ratio calculated for the same simulation as in Figure 1.

Figure: The behavior of common measures of association under a complex disease transmission pattern.

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Equations

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Biological efficacy

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# Hierarchical interpretation of common measures of efficacy

### TABLE 1. Parameters used for measuring various effects of vaccination\*

		Co	mparison groups and effe				
Level	Parameter choice	Susceptibility	Infectiousness	Combined change in susceptibility and infectiousness			
	Conditional on exposure to infection:	ak a,					
I	Transmission probability, p Secondary attack rate (SAR)	$VE_{s,p}\dagger = 1 - \frac{\rho_{01}}{\rho_{00}}$	$VE_{l,p} = 1 - \frac{\rho_{10}}{\rho_{00}}$	$VE_{r,p} = 1 - \frac{\rho_{11}}{\rho_{00}}$			
		Study design					
		l	IIA indirect	IIB total	III overall		
	Unconditional:						
п	Incidence rate (IR)	$VE_{_{SIR}} \approx 1 - \frac{IR_{_{A1}}}{IR_{_{A0}}}$	$VE_{_{IIA,ER}} = 1 - \frac{IR_{A0}}{IR_{B0}}$	$VE_{_{\mathrm{HB,HR}}} = 1 - \frac{IR_{A1}}{IR_{B0}}$	$VE_{_{\rm HUR}} = 1 - \frac{fIR_{\rm A1} + (1 - f)IR_{\rm A0}}{IR_{\rm B0}}$		
	Hazard, $\lambda$	$VE_{_{S,\lambda}}=1-\frac{\lambda_{A1}}{\lambda_{A0}}$	$VE_{_{UA,\lambda}} = 1 - \frac{\lambda_{A0}}{\lambda_{B0}}$	$VE_{_{NB,\lambda}} = 1 - \frac{\lambda_{A1}}{\lambda_{B0}}$	$VE_{_{W,\lambda}} = 1 - \frac{f\lambda_{A1} + (1 - f)\lambda_{A0}}{\lambda_{B0}}$		
ш	Proportional hazards (PH)	$VE_{_{\mathcal{S}^{PH}}} = 1 - e^{\beta_1}$	NAţ	NA	NA		
IV	Cumulative incidence (CI)	$VE_{g_{CI}} = 1 - \frac{CI_{A1}}{CI_{A0}}$	$VE_{HA,CI} = 1 - \frac{CI_{A0}}{CI_{B0}}$	$VE_{_{IB,CI}} = 1 - \frac{CI_{A1}}{CI_{B0}}$	$VE_{II,CI} = 1 - \frac{fCI_{A1} + (1 - f)CI_{A0}}{CI_{B0}}$		

# Figure: Interpretation of efficacy measures E Sage

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- Equations
- Glossary
- **Biological efficacy**
- Challenge
- Study Desi
- Rehavior
- Hierarchy
- Causal inference

# Causal inference for infectious diseases

