




Predicting and Controlling Future Zika Outbreaks

Eduardo Massad
School of Applied Mathematics
Fundacao Getulio Vargas
Brazil

Insert Logo of
eduardomassad@fqv.br


18/06/2018

Preparedness Latin American Network
ZikaPLAN



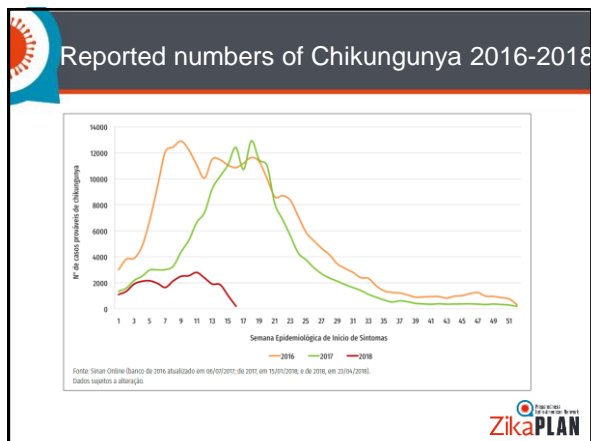
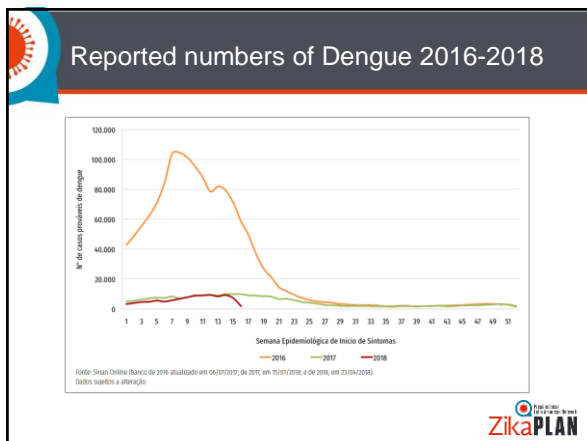
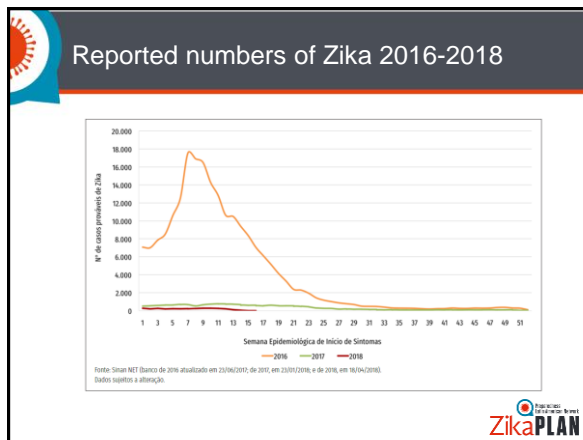
Has Zika burnt itself out in Brazil/Latin America?

ZikaPLAN



Will Zika resurge in the foreseeable future?

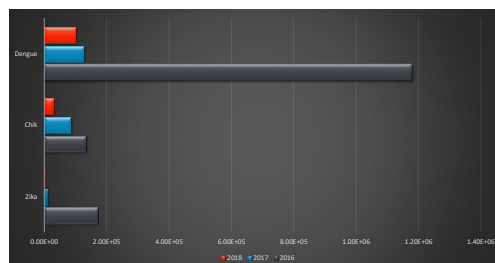
ZikaPLAN



Cases of Zika, Chik and Dengue, Brazil

	2016	2017	2018
Zika	170535	10286	2985
Chik	135030	86568	29675
Dengue	1.18E+06	128730	101863

Cases of Zika, Chik and Dengue, Brazil



Travellers from French Polynesia to Brazil

Table 1. Number of travellers departing French Polynesian airports with final destinations in Brazilian cities between December 2013 and February 2014

Departure Month	Departure Year	Destination City	Total Number
December	2013	Sao Paulo	20
December	2013	Rio de Janeiro	14
December	2013	Recife	11
December	2013	Porto Alegre	5
December	2013	Florianopolis	4
December	2013	Curitiba	4
January	2014	Sao Paulo	18
January	2014	Sao Paulo	16
January	2014	Goiânia	9
January	2014	Rio de Janeiro	9
January	2014	Curitiba	6
January	2014	Vitoria	7
January	2014	Fortaleza	2
January	2014	Rio de Janeiro	2
January	2014	Salvador	2
February	2014	Sao Paulo	20
February	2014	Porto Alegre	10
February	2014	Rio de Janeiro	9
February	2014	Curitiba	6
February	2014	Sao Paulo	4
February	2014	Uberlandia	2

ZIKV Infected Travellers Arriving in Brazil

Table 2. Individual risk of ZIKV infection in French Polynesia and the expected number of ZIKV infections exported to Brazil via travellers on commercial flights between December 2013 and February 2014

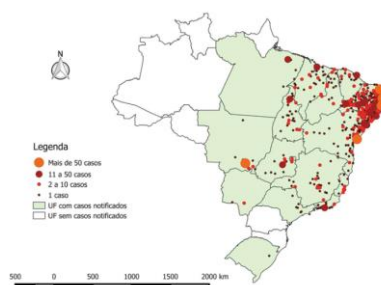
Week	Individual Probability of Infection	Expected Number of Travellers Infected with ZIKV
02/December/2013	0.0979584	3
09/December/2013	0.114501	2
16/December/2013	0.121446	1
23/December/2013	0.11955	0
30/December/2013	0.112717	4
07/January/2014	0.100661	2
14/January/2014	0.0845351	1
21/January/2014	0.068227	0
28/January/2014	0.0532315	2
04/February/2014	0.0407119	1
11/February/2014	0.0312288	0
18/February/2014	0.0227581	0
25/February/2014	0.0166078	1
Total		18

Theoretical R_0 in Brazilian Cities

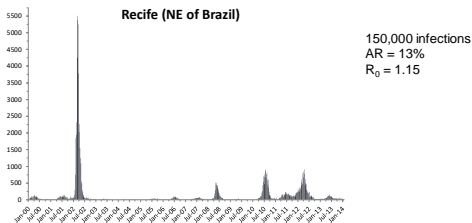
Table 3. Estimated Basic Reproduction number R_{0-ZIKV} for the selected Brazilian cities

	Expected number of ZIKV infected travelers						
	Rio de Janeiro	São Paulo	Recife	Fortaleza	Salvador	Goiânia	Vitoria
Dec/2013	2	2	1	0	0	0	0
Jan/2014	1	3	0	0	0	1	1
Feb/2014	0	1	0	0	0	0	0
Total	3	6	1	0	0	1	1
R_{0-ZIKV}	1.4	2.15	1.31	1.45	1.6	1.46	1.4

ZIKV Cases in Brazil, 2015

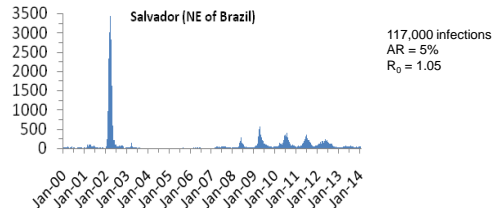


Dengue in Recife 2002



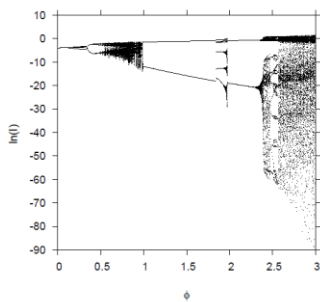
ZikaPLAN

Dengue in Salvador 2002



ZikaPLAN

The Chaotic Dynamics of Dengue



ZikaPLAN

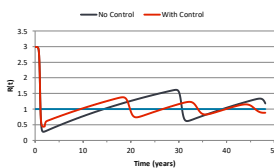
For vector-borne infections

$$R(t) = \frac{ma^2 bce^{-\mu_M \tau}}{\mu_M (\mu_H + \gamma_H + \alpha_H)} \frac{S(t)}{N(t)}$$

ZikaPLAN

Effective Reproductive Number $R(t)$

$$R(t) = \frac{ma^2 bce^{-\mu_M \tau}}{\mu_M (\mu_H + \gamma_H + \alpha_H)} \frac{S_H(t)}{N_H(t)}$$



ZikaPLAN

The Ross-Macdonald Model

Humans

$$\frac{dS_H}{dt} = -abI_H \frac{S_H}{N_H} + \mu_H (N_H - S_H)$$

$$\frac{dI_H}{dt} = abI_H \frac{S_H}{N_H} - (\mu_H + \gamma_H) I_H$$

$$\frac{dR_H}{dt} = \gamma_H I_H - \mu_H R_H$$

Mosquitoes

$$\frac{dS_M}{dt} = -acS_H \frac{I_H}{N_H} + \mu_M (L_M + I_M) + \frac{dN_M}{dt}$$

$$\frac{dL_M}{dt} = acS_H \frac{I_H}{N_H} - \gamma_M I_M - \mu_M L_M$$

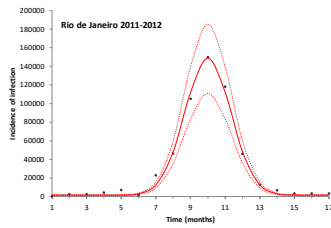
$$\frac{dI_M}{dt} = \gamma_M L_M - \mu_M I_M$$

$$N_H = S_H + I_H + R_H$$

$$N_M = S_M + L_M + I_M$$

ZikaPLAN

Fitting Dengue Incidence



ZikaPLAN

Incidence

$$Incidence_{DENV}(t) = c_1 \exp\left[-\frac{(t-c_2)^2}{c_3}\right] + c_4$$

$$Incidence(t) = \lambda(t)S_H(t) = ab \frac{I_M(t)}{N_H} S_H(t)$$

ZikaPLAN

Susceptible, Latent and Infected Mosquitoes

$$I_M(t) = \frac{Incidence_{DENV}(t)N_H(t)}{abS_H(t)}$$

$$L_M(t) = \frac{1}{\gamma_M} \left[\frac{d}{dt} I_M(t) + \mu_M I_M(t) \right]$$

$$S_M(t) = \frac{N_H}{a\lambda_H(t)} \left[\frac{d}{dt} L_M(t) + (\mu_M + \gamma_M)L_M(t) \right]$$

ZikaPLAN

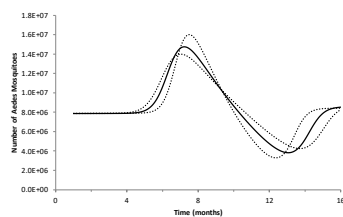
Total Number of Mosquitoes

$$N_M = S_M(t) + L_M(t) + I_M(t)$$

$$\frac{dN_M(t)}{dt} = \frac{dS_M(t)}{dt} + \frac{dL_M(t)}{dt} + \frac{dI_M(t)}{dt}$$

ZikaPLAN

Mosquitoes' Seasonality

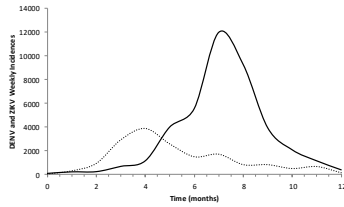


ZikaPLAN

Predicting Zika resurgence in the foreseeable future

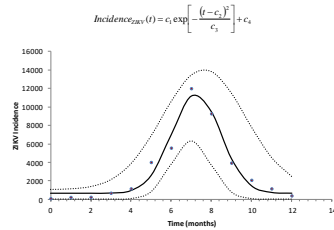
ZikaPLAN

ZIKV (continuous line) and Dengue (dotted line) Salvador 2015



ZikaPLAN

Fitting ZIKV Incidence Salvador 2015



ZikaPLAN

Number of Aedes mosquitoes Salvador 2015

$$Incidence_{ZIKV}(t) = c_1 \exp\left[-\frac{(t-c_2)^2}{c_3}\right] + c_4$$

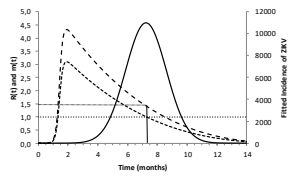
$$I_M(t) = \frac{Incidence_{ZIKV}(t) N_H(t)}{ab S_H(t)}$$

$$L_M(t) = \frac{1}{\gamma_M} \left[\frac{d}{dt} I_M(t) + \mu_M I_M(t) \right]$$

$$S_M(t) = \frac{N_H}{acI_H(t)} \left[\frac{d}{dt} L_M(t) + (\mu_M + \gamma_M) L_M(t) \right]$$

ZikaPLAN

Fitted ZIKV Incidence(continuous line), mosquitoes' density (gross dotted line) and effective reproduction number (fine dotted line), Salvador 2015



$$m_{crit}(t) \sim 1.5$$

ZikaPLAN

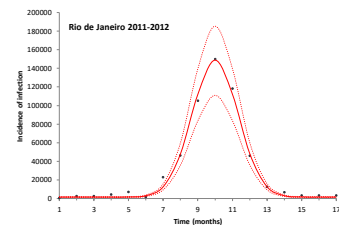
Critical density of mosquitoes, m_{crit}

$$m_{crit} = \frac{N_H}{S_H(t_{crit})} \kappa,$$

$$\kappa = \frac{\mu_M(\mu_H + \gamma_H)(\mu_M + \gamma_M)}{a^2 bc \gamma_M}$$

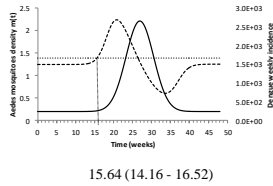
ZikaPLAN

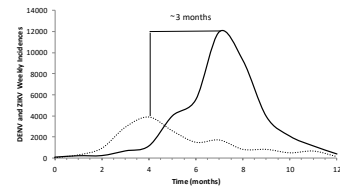
Fitting Dengue Incidence Rio de Janeiro - 2011/12

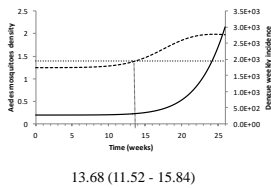


ZikaPLAN

Incidence and Density of Aedes mosquitoes



Timeline Gap Between ZIKV and Dengue
Salvador 2015

Incidence and Density of Aedes mosquitoes
in the first 25 weeks

Sexual Transmission



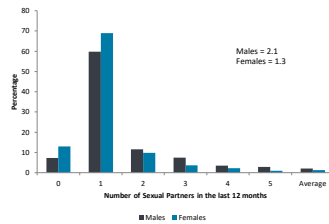
Basic Reproduction Number
of Sexual Transmission

$$R_0 = n\beta cT$$

Number of Sexual Partners

$$n \equiv \frac{\langle i^2 \rangle}{\langle i \rangle}$$

Brazilians sexual behaviour



ZikaPLAN

Brazilians sexual behaviour

	Males	Fem
0	7.2	12.9
1	59.8	69
2	11.6	9.8
3	7.5	3.7
4	3.5	2.2
5	2.9	0.9
m	2.1	1.3
σ^2	22	26

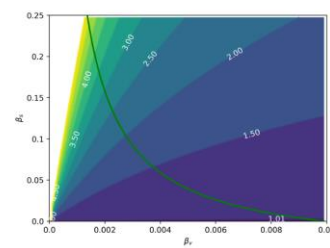
ZikaPLAN

Mean Number of sexual partners

$$n \equiv 2.1 + \frac{22}{2.1} = 10.4$$

ZikaPLAN

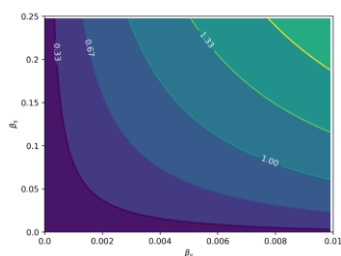
Attack rates ratios of females and males



Ratio ARW/ARM for a range of β_s and β_v values. The green line represents $R_0 = 1$, i.e. the epidemic threshold. Any point to the right of this curve, has $R_0 > 1$. It is worth noticing that the reported excess cases reported for Zika in women are possible both during epidemics and off-season.

ZikaPLAN

R_0 and relative transmission intensities – sexual vs vectorial



R_0 as a function of the relative intensities of sexual (β_s) and vectorial (β_v) transmissions. The R_0 values are already adjusted for the heterogeneity in sexual contact rates.

41

ZikaPLAN

Outbreak Vaccine

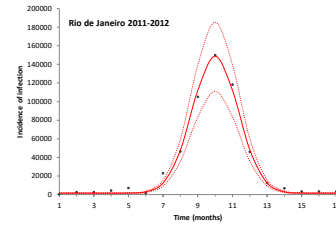
ZikaPLAN

The Model

$$\begin{aligned}
 \frac{dN_u(t)}{dt} &= r_u \left[1 - \frac{N_u(t)}{K_u} \right] \left[S_u(t) + R_u(t) + gI_u(t) \right] - \mu_u N_u(t) \\
 \frac{dS_u(t)}{dt} &= -abI_u(t) \frac{S_u(t)}{N_u(t)} - \mu_u S_u(t) + r_u \left[1 - \frac{N_u(t)}{K_u} \right] \left[S_u(t) + R_u(t) + gI_u(t) \right] - \\
 &\quad - (S_u(t)\theta(t) - I_u(t)\theta(t) - t) \\
 \frac{dI_u(t)}{dt} &= abI_u(t) \frac{S_u(t)}{N_u(t)} - (\mu_u + \gamma_u) I_u(t) \\
 \frac{dR_u(t)}{dt} &= \gamma_u I_u(t) - \mu_u R_u(t) \\
 \frac{dS_a(t)}{dt} &= (S_u(t)\theta(t) - I_u(t)\theta(t) - t) - \mu_a S_a(t) \\
 \frac{dI_a(t)}{dt} &= r_a \left[1 - \frac{N_a(t)}{K_a} \right] \left[(1 - \xi)I_u(t) \right] - pabI_u(t) - (1 - p)\gamma_a I_a(t) \\
 \frac{dR_a(t)}{dt} &= (1 - p)\gamma_a I_a(t) - \mu_a R_a(t) \\
 \frac{dM_a(t)}{dt} &= pabI_u(t) - (\mu_a + \alpha_a) M_a(t) \\
 \frac{dS_a(t)}{dt} &= -acS_a(t) \frac{[I_u(t) + I_a(t)]}{N_u(t)} + \mu_a [L_u(t) + L_a(t)] + \frac{dN_a}{dt} \\
 \frac{dL_u(t)}{dt} &= acS_a(t) \frac{[I_u(t) + I_a(t)]}{N_u(t)} - (\mu_u + \gamma_u) L_u(t) \\
 \frac{dL_a(t)}{dt} &= \gamma_u L_u(t) - \mu_a L_a(t) \\
 I_u^*(t) &= I_u^*(0) \exp[-(\mu_u + \gamma_u + \alpha_a)(t - t_1)] \theta(t - t_1) \\
 N_u(t) &= S_u(t) + L_u(t) + I_u(t)
 \end{aligned}$$



Fitting Dengue Incidence

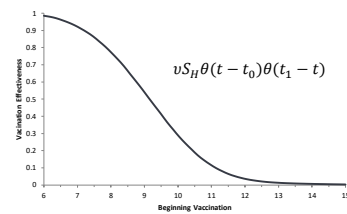


Number of Aedes Mosquitoes

$$\begin{aligned}
 Incidence_{ZIKV}(t) &= c_1 \exp\left[-\frac{(t - c_2)^2}{c_3}\right] + c_4 \\
 I_M(t) &= \frac{Incidence_{ZIKV}(t) N_H(t)}{abS_H(t)} \\
 L_M(t) &= \frac{1}{\gamma_M} \left[\frac{d}{dt} I_M(t) + \mu_M I_M(t) \right] \\
 S_M(t) &= \frac{N_H}{acI_H(t)} \left[\frac{d}{dt} L_M(t) + (\mu_M + \gamma_M) L_M(t) \right]
 \end{aligned}$$



Optimum moment, t_0 , to introduce the vaccine

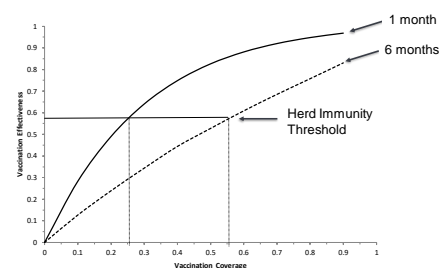


Definition of Vaccine Efficacy

$$\begin{aligned}
 Efficacy &= 1 - \frac{Cases_{after}}{Cases_{before}} \\
 Cases &= \int_0^{\infty} abI_M(t) \frac{S_H(t)}{N(t)} dt
 \end{aligned}$$



Optimum Duration of the Campaign

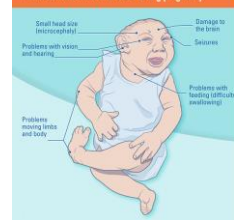


Routine Vaccine

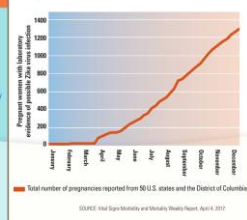


Congenital Zika Syndrome

Congenital Zika syndrome is a pattern of birth defects in babies infected with Zika during pregnancy

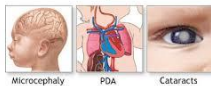


Reported cases of pregnant women with any lab evidence of possible Zika increased in 2016



Congenital Rubella Syndrome

Rubella syndrome



©ADAM



Lifetime Expected Risk of Infection

The sum over all ages of the product of the force of infection (per capita incidence rate, $\lambda(a)$, times the seroconversion rate, $C(a)$, times the disease's "undesirability", $R(a)$, times the probability of infected children not seroconverting.



Lifetime Expected Risk of Infection

The sum over all ages of the product of the force of infection (per capita incidence rate, $\lambda(a)$, times the seroconversion rate, $C(a)$, times the disease's "undesirability", $R(a)$, times the probability of infected children not seroconverting.

$$E = \int_0^{\infty} \lambda(a) C(a) R(a) \exp \left[- \int_0^a \lambda(a') C(a') da' \right] da$$



Lifetime Expected Risk of Infection

$$E = \int_0^{\infty} \lambda(a) C(a) R(a) \exp \left[- \int_0^a \lambda(a') C(a') da' \right] da$$



Lifetime Expected Risk of Infection

Force of Infection (Incidence)

$$E = \int_0^{\infty} \lambda(a)C(a)R(a) \exp\left[-\int_0^a \lambda(a')C(a')da'\right] da$$



Lifetime Expected Risk of Infection

Seroconversion Rate of the Vaccine ("Taking")

$$E = \int_0^{\infty} \lambda(a)C(a)R(a) \exp\left[-\int_0^a \lambda(a')C(a')da'\right] da$$



Lifetime Expected Risk of Infection

"Undesirability" of the infection (with age)

$$E = \int_0^{\infty} \lambda(a)C(a)R(a) \exp\left[-\int_0^a \lambda(a')C(a')da'\right] da$$



Lifetime Expected Risk of Infection

Probability of infected children not seroconverting.

$$E = \int_0^{\infty} \lambda(a)C(a)R(a) \exp\left[-\int_0^a \lambda(a')C(a')da'\right] da$$



Seroprevalence of Infection

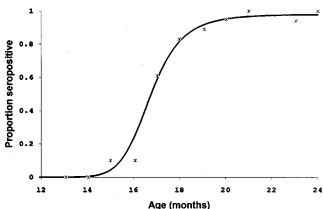


FIGURE 4 Seroprevalence obtained between 12 and 24 months of age (stars), fitted to a logistic function (line), showing the seroconversion attributed to the routine vaccination programme



Calculating the Force of Infection

$$\lambda(a) = \frac{[-dS^+(a)]/da}{[1 - S^+(a)]}$$



Average Age of First Infection

$$\bar{A}_{1st} = \frac{\int_0^{\infty} a \lambda(a) S_H(a) da}{\int_0^{\infty} \lambda(a) S_H(a) da} \frac{1}{\lambda}$$



“Typical” form of the Force of Infection

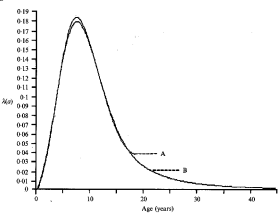


Fig. 5. Age-dependent force of infection, $\lambda(a)$, estimated by catalytic methods (curve A) through the equation, described in [14], [15]:
 $\lambda(a) = \frac{[-dS^*(a)]}{[1 - S^*(a)]} da$
where $S^*(a)$ is the age-dependent proportion of seropositives, as compared with that estimated by applying equation (15) (curve B).



Impact of Vaccination on $\lambda(a)$

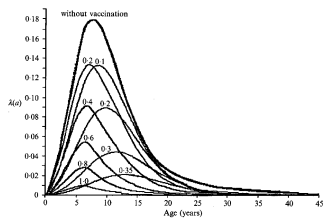


Fig. 6. Impact of two alternative vaccination strategies on the force of infection. The set of curves under that without vaccination represents the force of infection after the vaccination between 1 and 2 years of age (right curves labelled with coverage proportions 0.1-0.35), and between 7 and 8 years of age (left curves labelled with coverage proportions 0.2-1.0).



Fertility curve as a proxy for $R(a)$

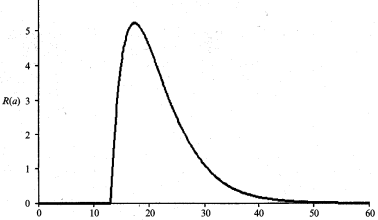


Fig. 2. Women reproductive function, $R(a)$, as fitted by equation 16 to demographic data from the State of São Paulo, Brazil.



Optimum Age to Vaccinate

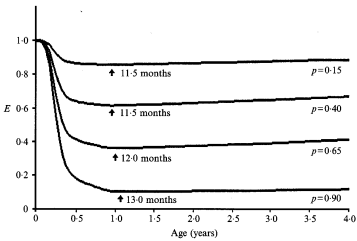


Fig. 3. Lifetime Expected Risk of acquisition of rubella, E , as a function of the age of vaccination and proportion of coverage in the routine vaccination scheme chosen.




Relative Impact on Congenital Syndrome

The ratio of the number of congenital syndrome cases after vaccination over the the number of congenital syndrome Cases before vaccination.


$$\rho = \frac{\int_0^a r_p(a') \lambda_p(a') S_p(a') da'}{\int_0^a r_b(a') \lambda_b(a') S_b(a') da'}$$






Relative Impact on Congenital Syndrome

$$\rho = \frac{\int_0^a r_p(a') \lambda_p(a') S_p(a') da'}{\int_0^a r_b(a') \lambda_b(a') S_b(a') da'} da$$







Relative Impact on Congenital Syndrome

Incidence of CS in Infected Pregnant Women Before Vaccination

$$\rho = \frac{\int_0^a r_p(a') \lambda_p(a') S_p(a') da'}{\int_0^a r_b(a') \lambda_b(a') S_b(a') da'} da$$







Relative Impact on Congenital Syndrome

Incidence of CS in Infected Pregnant Women Before Vaccination

$$\rho = \frac{\int_0^a r_p(a') \lambda_p(a') S_p(a') da'}{\int_0^a r_b(a') \lambda_b(a') S_b(a') da'} da$$

Incidence of CS in Infected Pregnant Women After Vaccination





Relative Impact on Congenital Syndrome

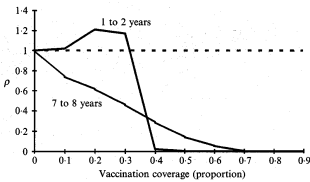




Fig. 8. Relative number of cases of CRS after vaccination as compared with before vaccination as a function of coverage proportions. Broken line denoting 1 indicates no effect of intervention, whilst continuous lines represent the impact of vaccination between 1 and 2 years of age, as compared with vaccination between 7 and 8 years of age. It can be noted that, for low coverages, the latter strategy is more effective. On the other hand, for high coverages, the former strategy has a greater impact on the CRS incidence.







Impact of Vaccination on CRS

TABLE 4 Reported number of cases of congenital rubella syndrome

	1992	1993	1994
Confirmed	16	1	0
Suspected	13	5	0
Discarded	18	8	1
Total	48	15	1





“Prediction is very difficult, especially about the future”

Niels Bohr

