

Modeling infectious diseases

Modeling infectious diseases using R: Practical Session

Static and Dynamic aspects of SIR model

What do we cover in this practical session ?

- Simple transmission models in R.
- SIR: static aspect (time homogeneity).
- SIR: dynamic aspects.
- Vaccination.
- Software: the **deSolve** package in R.

R requirements

- Basic knowledge in R.
 - Programming of a user defined simple function.
-
- Why R ?
 1. Free.
 2. Fast (not in our case).
 3. Updated.
 4. Documented.

Part 1:
time homogeneity

SIR model :time homogeneity

$$\begin{aligned}\frac{dS(a)}{da} &= -\lambda S(a) \\ \frac{dI(a)}{da} &= \lambda S(a) - \sigma I(a) \\ \frac{dR(a)}{da} &= \sigma I(a)\end{aligned}$$

The force of infection: the rate in which individuals are infected

The recovery rate: the rate in which individuals recovered and move to the immune class.
Assumption: life long immunity.

SIR model: time homogeneity

$$\begin{aligned}\frac{dS(a)}{da} &= -\lambda S(a) \\ \frac{dI(a)}{da} &= \lambda S(a) - \sigma I(a) \\ \frac{dR(a)}{da} &= \sigma I(a)\end{aligned}$$

The force of infection = 0.2.
On average: 5 years in the susceptible class.

The recovery rate: 10 days.

The unit of the parameters are in years

Transmission models in R

- We need to integrate the system of the ordinary differential equation.
- deSolve package in R.
- Numerical integration using of ODE system.

$$\begin{aligned}\frac{dS(a)}{da} &= -\lambda S(a) \\ \frac{dI(a)}{da} &= \lambda S(a) - \sigma I(a) \\ \frac{dR(a)}{da} &= \sigma I(a)\end{aligned}$$

Transmission models in R

SIR model

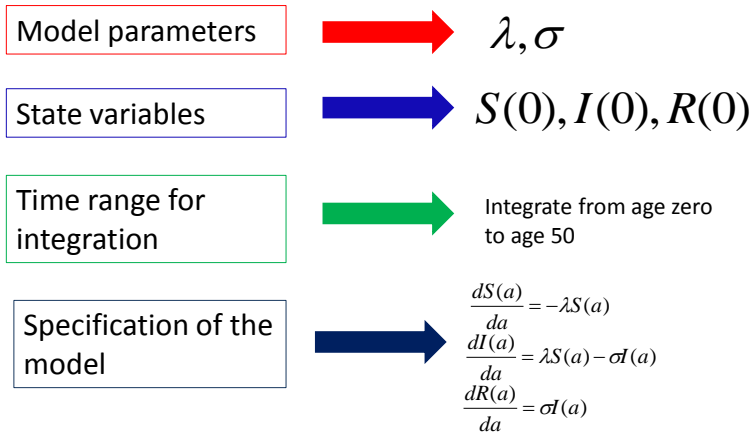
$$\begin{aligned}\frac{dS(a)}{da} &= -\lambda S(a) \\ \frac{dI(a)}{da} &= \lambda S(a) - \sigma I(a) \\ \frac{dR(a)}{da} &= \sigma I(a)\end{aligned}$$

Specification in R

- Model parameters.
- State variables (the value of the parameters at age (time) zero).
- Time range (=age range) for integration.

Transmission models in R

R program



Specification of the model parameters in R

```
> parameters <- c(lambda = 0.2, v=36.5)
> parameters
lambda      v
0.2      36.5
```

The force of infection = 0.2 years.

$$5^{-1} = 0.2$$

Recovery rate 10 days.

$$\left(\frac{10}{365}\right)^{-1} = 36.5$$

The unit of the parameters are in years⁻¹

The state variables (initial values at age 0)

•Let us assume that the cohort size is 5000.

•At age=0:

$$S(0) = 4999$$

$$I(0) = 1$$

$$R(0) = 0$$

•Specification in R:

```
> state <- c(X=4999,Y=1,Z=0)
> state
  X  Y  Z
4999 1  0
```

Specification of the model in R

$$\left. \begin{aligned} \frac{dS(a)}{da} &= -\lambda S(a) \\ \frac{dI(a)}{da} &= \lambda S(a) - \sigma I(a) \\ \frac{dR(a)}{da} &= \sigma I(a) \end{aligned} \right\}$$

```
SIR<-
function(t,state,parameters)
{
  with(as.list(c(state,
    parameters)),
    {
      dX <- -lambda*X
      dY <- lambda*X - v*Y
      dZ <- v*Y
      list(c(dX, dY, dZ))
    })
}
```

We ask from the function to return the values of S, I and R

Specification of the time units for the integration

$$\begin{aligned}\frac{dS(a)}{da} &= -\lambda S(a) \\ \frac{dI(a)}{da} &= \lambda S(a) - \sigma I(a) \\ \frac{dR(a)}{da} &= \sigma I(a)\end{aligned}$$

- The solution of the model: numerical integration.
- Time units: age.
- Integration from age 0 to age 40 by unit of 0.01 years

```
> times<-seq(0,40,by=0.01)
> times
 [1] 0.00 0.01 0.02 0.03 0.04 0.05 0.06 0.07 0.08 0.09 0.10 0.11
[13] 0.12 0.13 0.14 0.15 0.16 0.17 0.18 0.19 0.20 0.21 0.22 0.23
[25] 0.24 0.25 0.26 0.27 0.28 0.29 0.30 0.31 0.32 0.33 0.34 0.35
[37] 0.36 0.37
```

Running the model

```
require(deSolve)
out<-as.data.frame(ode(y=state,times=times,func=SIR,parms=parameters))
```

The state variables:
the values at age 0.

The model
parameters: force of
infection (0.2) and
recovery rate (10
days)

Running the model

```
require(deSolve)
out<-as.data.frame(ode(y=state,times=times,func=SIR,parms=parameters))
```

Time range for
integration.

The model

Numerical integration
using ordinary
differential equation

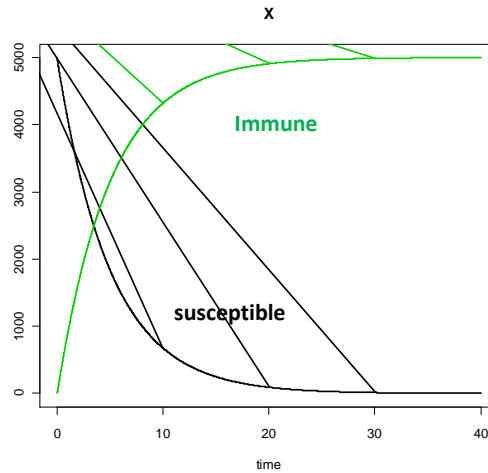
Solution

```
> require(deSolve)
> out <-
as.data.frame(ode(y=state,times=times,func=SIR,parms=
parameters))
> head(out)
```

	time	X	Y	Z
1	0.00	4999.000	1.000000	0.000000
2	0.01	4989.012	9.061818	1.926190
3	0.02	4979.044	14.641580	6.314481
4	0.03	4969.096	18.498345	12.405853
5	0.04	4959.168	21.159066	19.673391
6	0.05	4949.259	22.989501	27.751380

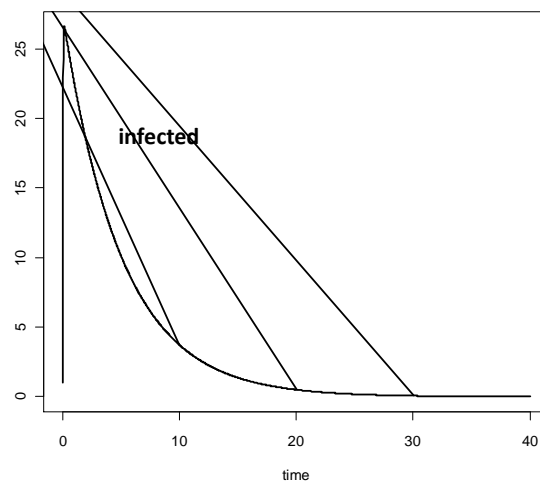
Graphical output

```
> plot (times,out$X ,type="l",main="X", xlab="time", ylab="-",lwd=2)
> lines(times,out$Z,col=3,lwd=2)
```

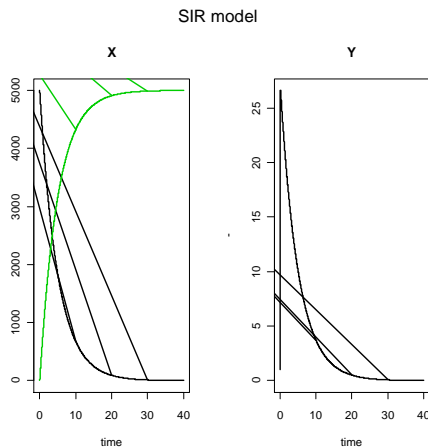


Graphical output

```
> plot (times,out$Y ,type="l",main="Y", xlab="time", ylab="-",lwd=2)
```



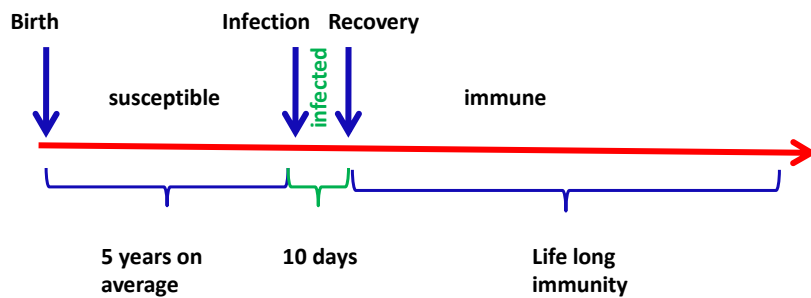
Graphical output



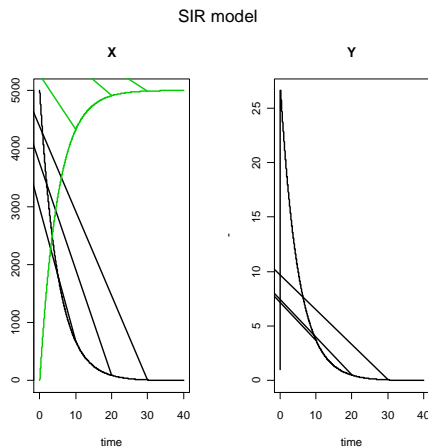
What do we see here ?

Number of infected individuals at each age.

Duration of stay in the different compartments of the models



Graphical output

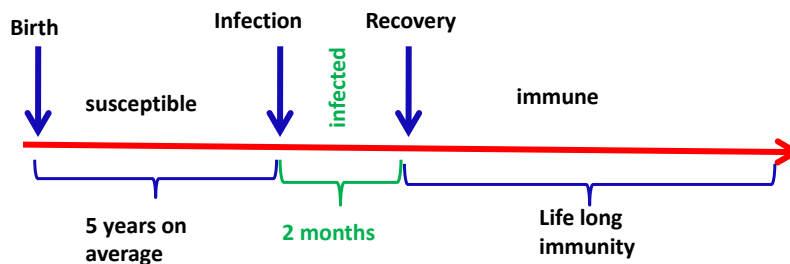


We expect to see only few infected individuals at each age (compared to the number of susceptible and immune).

Duration of stay in the different compartments of the models

Let us assume that the recovery rate is 2 months (i.e. individuals stay in the infected class 2 months)

What do we expect to see ?



The model with recovery rate of 2 months

$1/(60/365)$

parameters <- c(lambda = 0.2, v=**6.083333**)
parameters

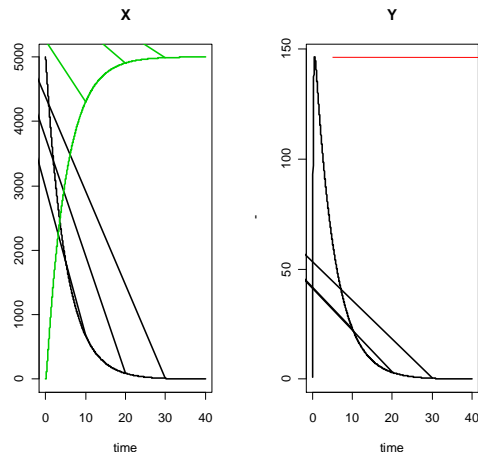
$$\left(\frac{60}{365}\right)^{-1} = 6.08$$

```
state <- c(X=4999,Y=1,Z=0)
state
SIR<-function(t,state,parameters)
{
  with(as.list(c(state, parameters)),
  {
    dX <- -lambda*X
    dY <- lambda*X - v*Y
    dZ <- v*Y
    list(c(dX, dY, dZ))
  })
}
times<-seq(0,40,by=0.01)
times
require(deSolve)
out <- as.data.frame(ode(y=state,times=times,func=SIR,parms=parameters))
head(out)
par(mfrow=c(1,2), oma=c(0,0,3,0))
plot(times,out$X ,type="l",main="X", xlab="time", ylab="",lwd=2)
lines(times,out$Z,col=3,lwd=2)
plot(times,out$Y ,type="l",main="Y", xlab="time", ylab="",lwd=2)
mtext(outer=TRUE,side=3,"SIR model",cex=1.5)
```

Recover rate of 2 months
(60 days)

Graphical output (force of infection of 0.2 and recovery rate of 2 months)

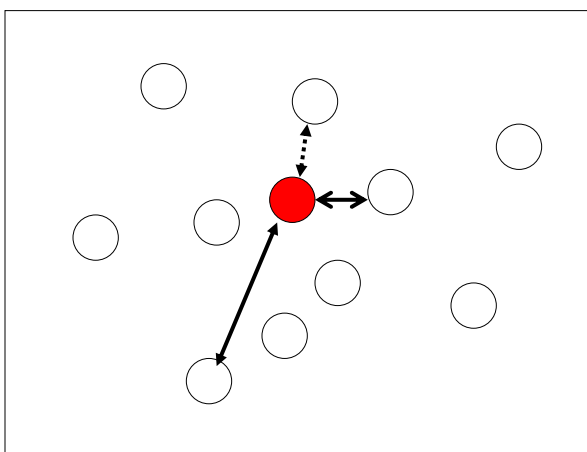
SIR model



~ 150 infected individuals at the peak (compare with ~25 for recovery rate of 10 days)

Part 2: Mass action principle

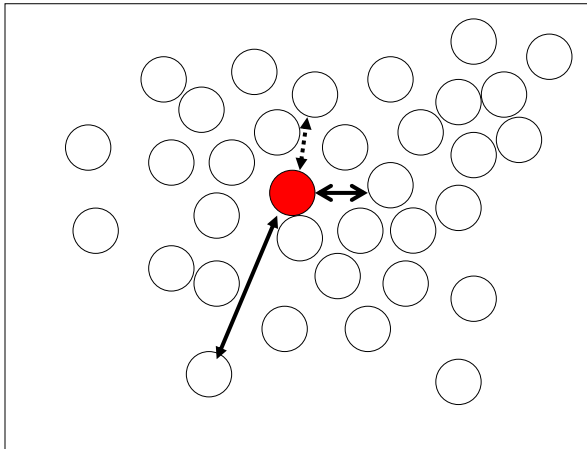
The Mass-Action Principle



Contacts are made in random.

Number of new cases = $P(\text{transmission}) \times \# \text{ of infectious} \times \# \text{ of susceptible}$

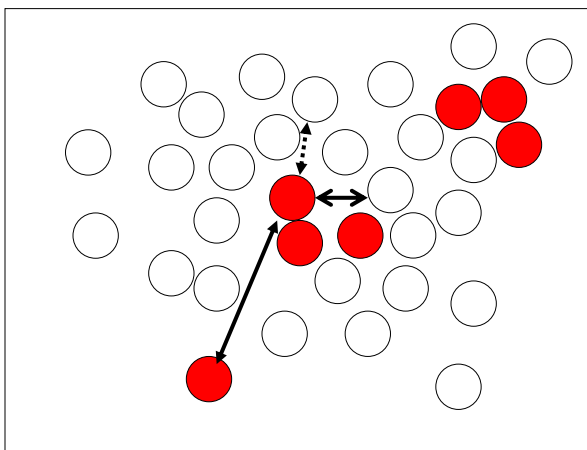
The Mass-Action Principle



Contacts are made in random.

Number of new cases = $P(\text{transmission}) \times \text{\# of infectious} \times \text{\# of susceptible}$

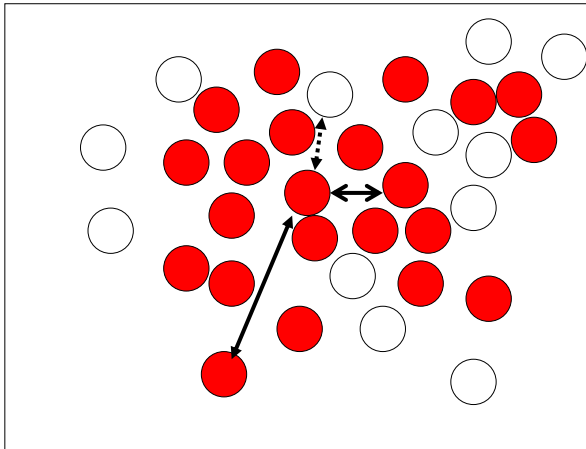
The Mass-Action Principle



Contacts are made in random.

Number of new cases = $P(\text{transmission}) \times \text{\# of infectious} \times \text{\# of susceptible}$

The Mass-Action Principle



Contacts are made in random.

Number of new cases:

$$\beta \times I \times S$$

Transmission
probability per
contact

Number of new cases = $P(\text{transmission}) \times \# \text{ of infectious} \times \# \text{ of susceptible}$

The Mass-Action Principle

$$\begin{array}{l}
 \frac{dS(a)}{da} = -\lambda S(a) \\
 \frac{dI(a)}{da} = \lambda S(a) - \sigma I(a) \\
 \frac{dR(a)}{da} = \sigma I(a)
 \end{array}
 \quad \rightarrow \quad
 \begin{array}{l}
 \frac{dS(a)}{da} = -\beta \times I(a) \times S(a) \\
 \frac{dI(a)}{da} = \beta \times I(a) \times S(a) - \sigma I(a) \\
 \frac{dR(a)}{da} = \sigma I(a)
 \end{array}$$

Age dependent force of infection

$$\begin{array}{lcl}
 \frac{dS(a)}{da} = -\lambda(a)S(a) & \xrightarrow{\quad} & \frac{dS(a)}{da} = -\beta \times I(a) \times S(a) \\
 \frac{dI(a)}{da} = \lambda(a)S(a) - \sigma I(a) & & \frac{dI(a)}{da} = \beta \times I(a) \times S(a) - \sigma I(a) \\
 \frac{dR(a)}{da} = \sigma I(a) & \xleftarrow{\quad} & \frac{dR(a)}{da} = \sigma I(a)
 \end{array}$$

$\lambda(a) = \beta \times I(a) \times S(a)$

↙ Age dependent ↓ constant

Model parameters

$$\begin{array}{l}
 \frac{dS(a)}{da} = -\beta \times I(a) \times S(a) \\
 \frac{dI(a)}{da} = \beta \times I(a) \times S(a) - \sigma I(a) \\
 \frac{dR(a)}{da} = \sigma I(a)
 \end{array}$$

$$\begin{array}{l}
 \beta = 0.0085 \\
 \sigma = 36.5 \quad (10 \text{ days})
 \end{array}$$

```

> parameters <- c(beta=0.0085, v=36.5)
> state <- c(X=4999, Y=1, Z=0)
> parameters
   beta      v 
0.0085 36.5000 
> state
   X   Y   Z 
4999  1   0
  
```

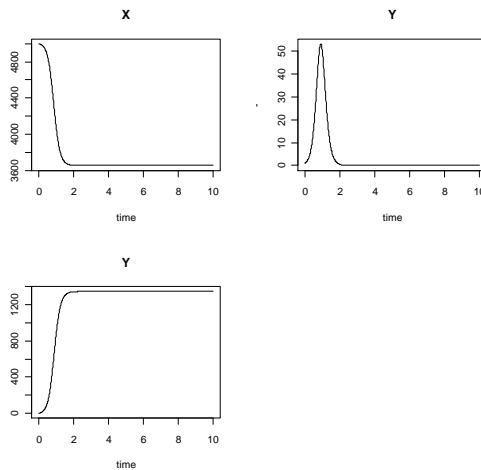

Specification of the model in R

$$\begin{aligned}\frac{dS(a)}{da} &= -\beta \times I(a) \times S(a) \\ \frac{dI(a)}{da} &= \beta \times I(a) \times S(a) - \sigma I(a) \\ \frac{dR(a)}{da} &= \sigma I(a)\end{aligned}$$

```
SIR<-function(t,state,parameters)
{
  with(as.list(c(state, parameters)),
  {
    dX <- -beta*Y*X
    dY <- beta*Y*X - v*Y
    dZ <- v*Y
    list(c(dX, dY, dZ))
  })
}
```

Numerical solution

SIR model

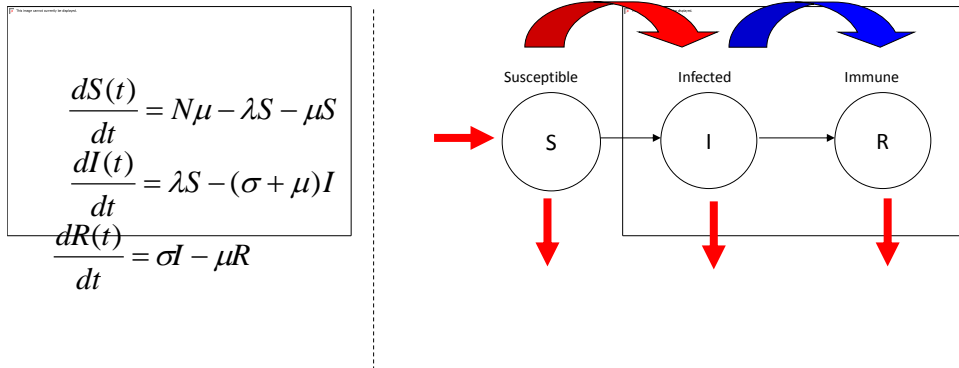


After ~ 2 years there
are no infected
individuals.

~3600 individuals will
not be infected.

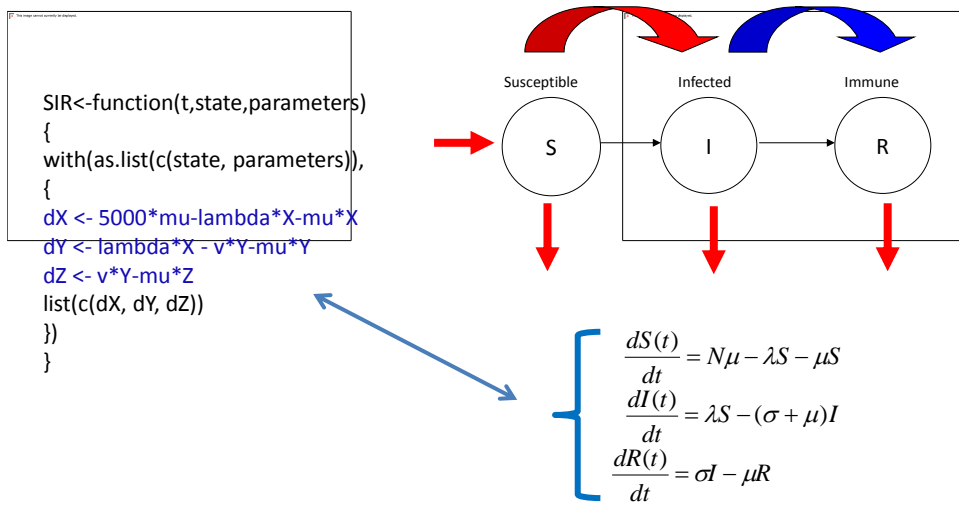
Equilibrium

SIR transmission model in open population



Birth and death rate are equal to μ (constant population size).

SIR transmission model in open population

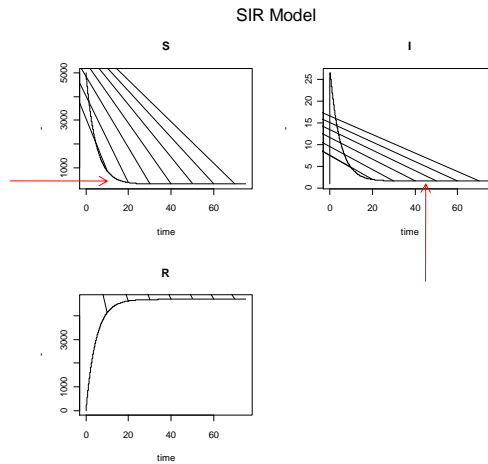


Open population (1)

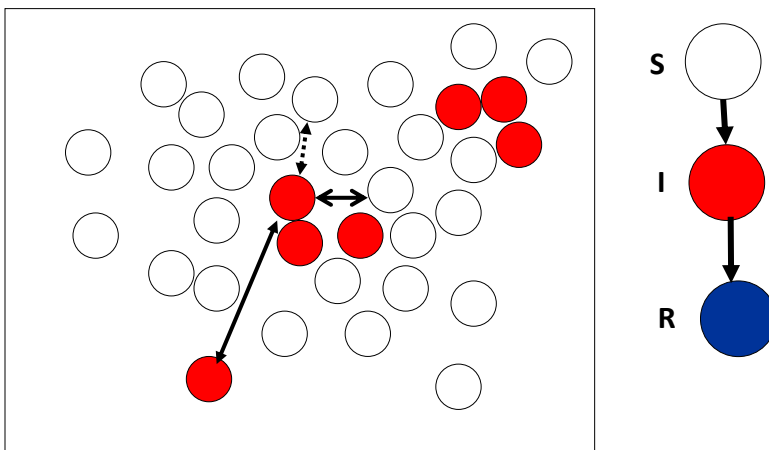
The model

$$\begin{aligned}\frac{dS(t)}{dt} &= N\mu - \lambda S - \mu S \\ \frac{dI(t)}{dt} &= \lambda S - (\sigma + \mu)I \\ \frac{dR(t)}{dt} &= \sigma I - \mu R\end{aligned}$$

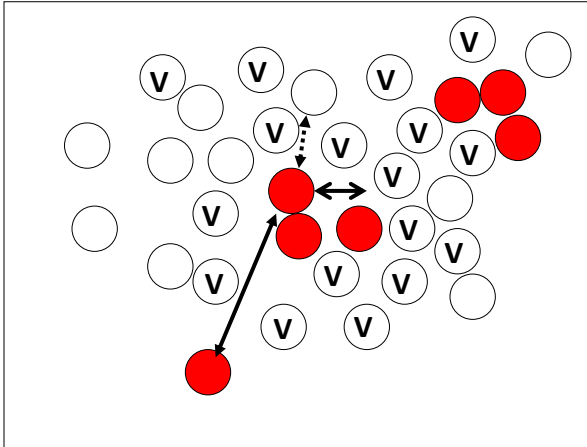
Equilibrium values



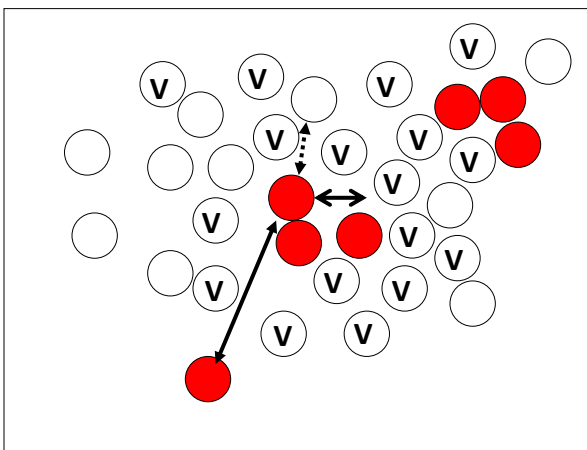
Vaccination



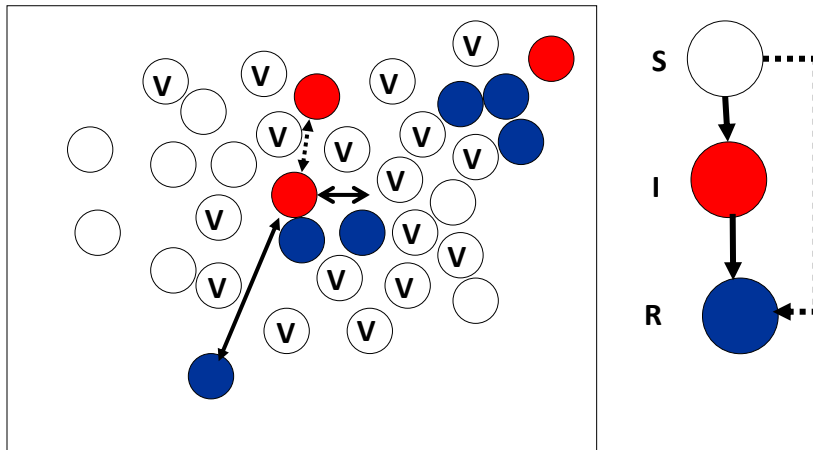
Vaccination



Vaccination

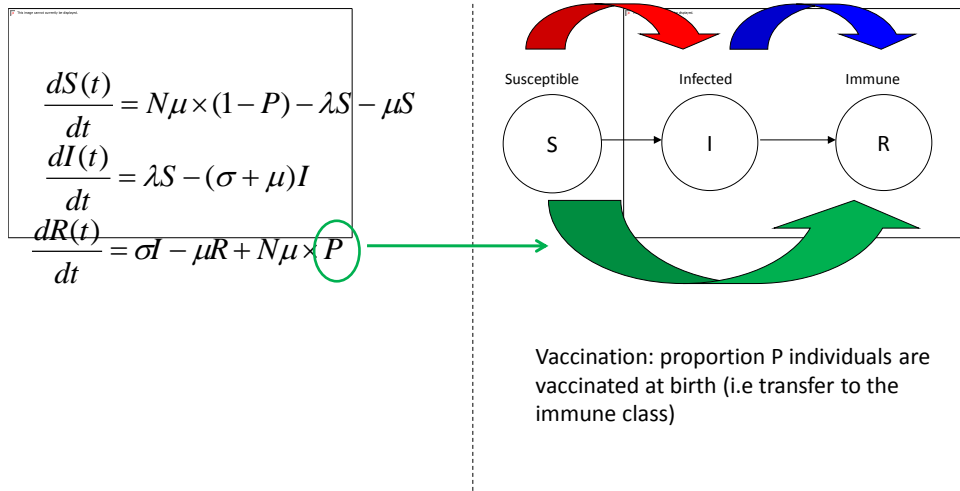


Vaccination

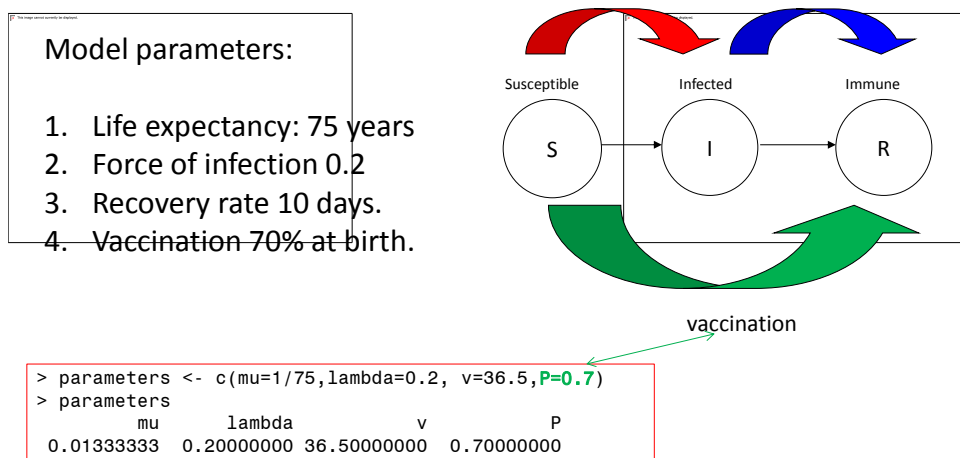


Part 3: vaccination in SIR model

Transmission model with vaccination



SIR transmission Model with vaccination

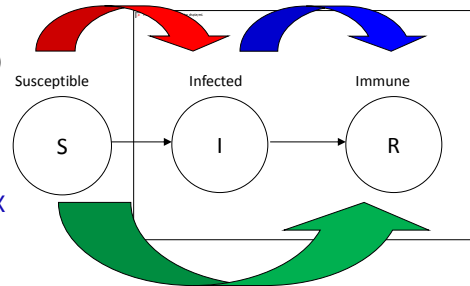


SIR transmission model with vaccination

```

SIR<-function(t,state,parameters)
{
  with(as.list(c(state,
  parameters))),
  {
    dX <- 5000*mu*(1-P)-lambda*X-mu*X
    dY <- lambda*X - v*Y-mu*Y
    dZ <- v*Y-mu*Z+5000*mu*P
    list(c(dX, dY, dZ))
  })
}

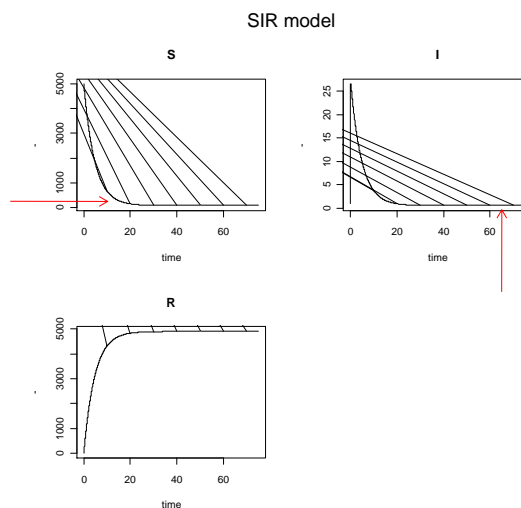
```



$$\begin{cases} \frac{dS(t)}{dt} = N\mu \times (1-P) - \lambda S - \mu S \\ \frac{dI(t)}{dt} = \lambda S - (\sigma + \mu)I \\ \frac{dR(t)}{dt} = \sigma I - \mu R + N\mu \times P \end{cases}$$

SIR transmission Model with vaccination

Equilibrium values



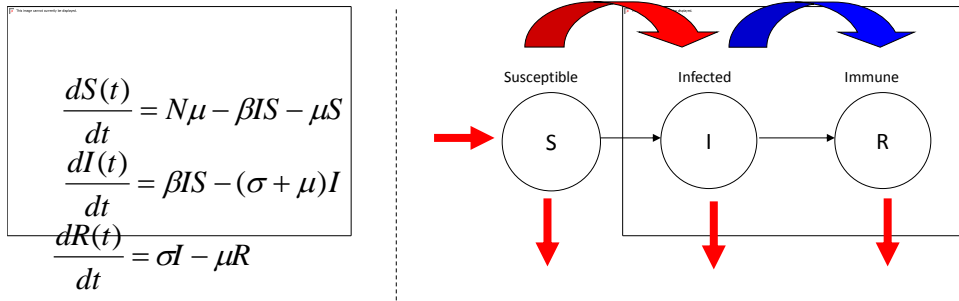
Part 4: Dynamic

SIR model in open population: Dynamic aspects

- In order to understand the dynamic of the SIR model we need to allow for time dependent force of infection.
- Open population.
- Mass action principle.

$$\lambda(t) = \beta \times I(t) \times S(t)$$

SIR transmission model in open population



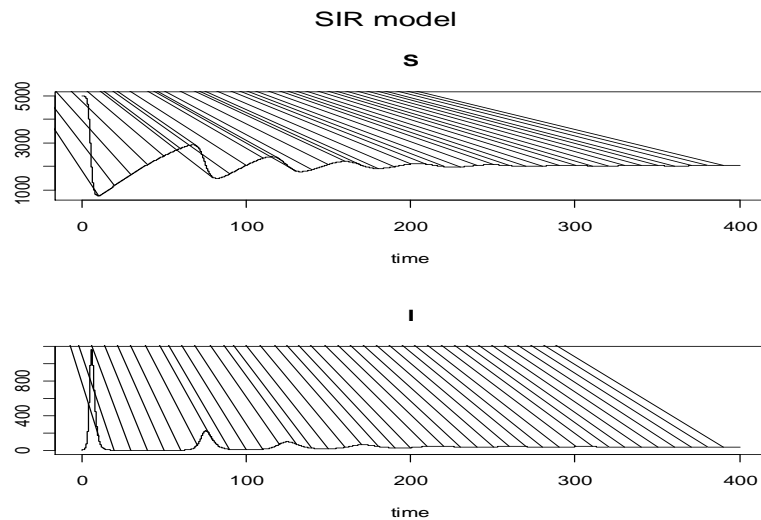
SIR transmission model in open population

$$\begin{aligned}\frac{dS(t)}{dt} &= N\mu - \beta IS - \mu S \\ \frac{dI(t)}{dt} &= \beta IS - (\sigma + \mu)I \\ \frac{dR(t)}{dt} &= \sigma I - \mu R\end{aligned}$$

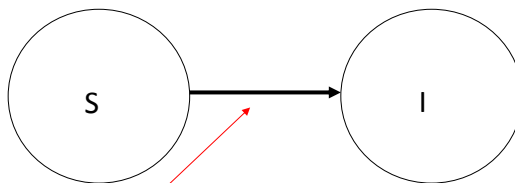
```

SIR<-function(t,state,parameters)
{
  with(as.list(c(state,
    parameters)),
    {
      dX <- 5000*mu-beta*Y*X - mu*X
      dY <- beta*Y*X - v*Y - mu*Y
      dZ <- v*Y -mu*Z
      list(c(dX, dY, dZ))
    })
}
```

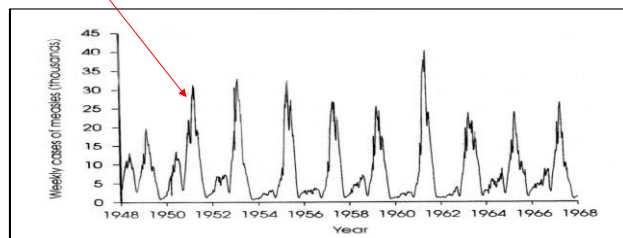
Solution for the model



Incidence data : Measles in UK



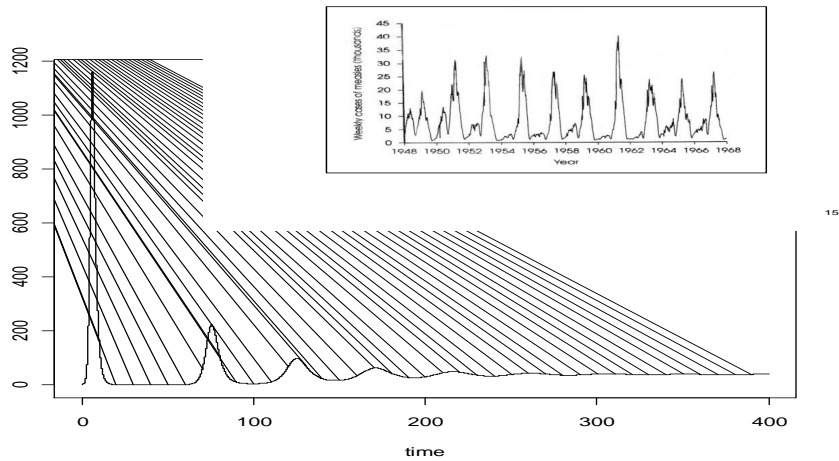
The **weekly** number of individuals who move from the susceptible to the infected class



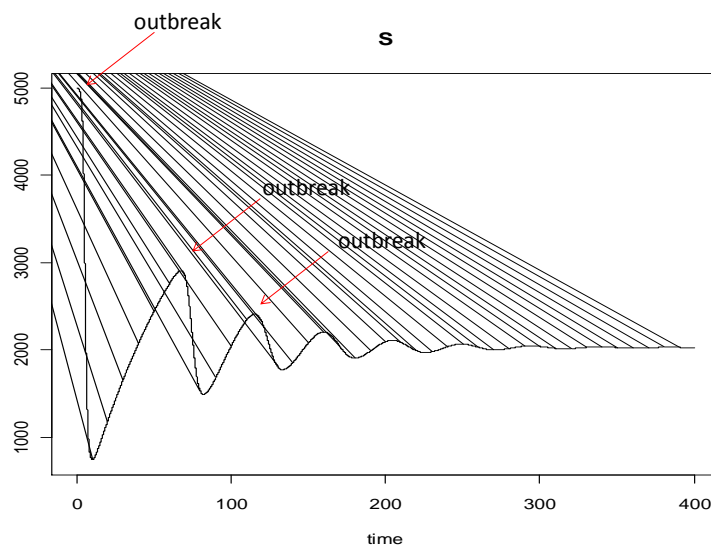
Time unit=week

15

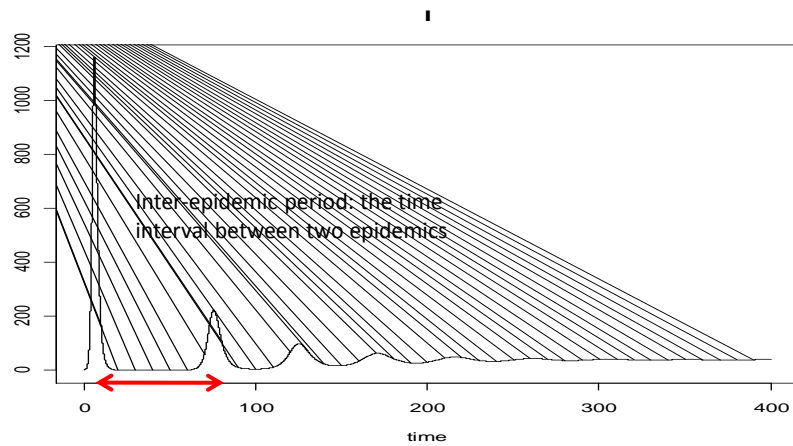
Observed outbreak and predicted outbreak



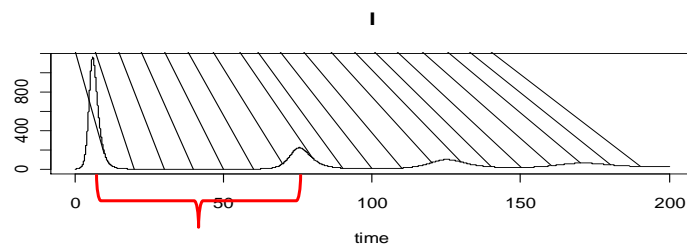
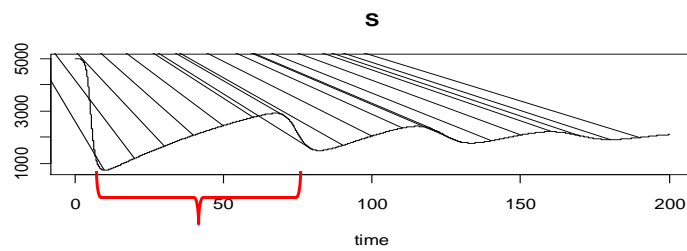
The susceptible class



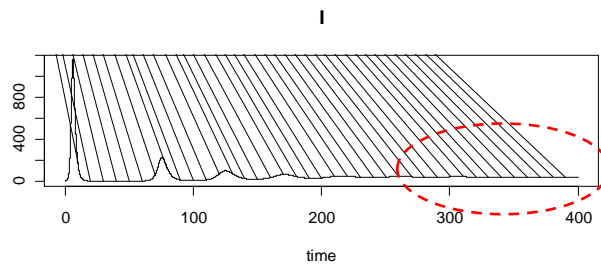
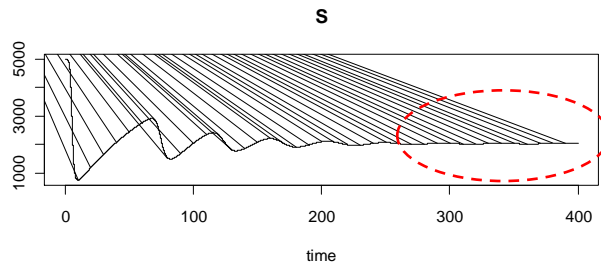
The infected class



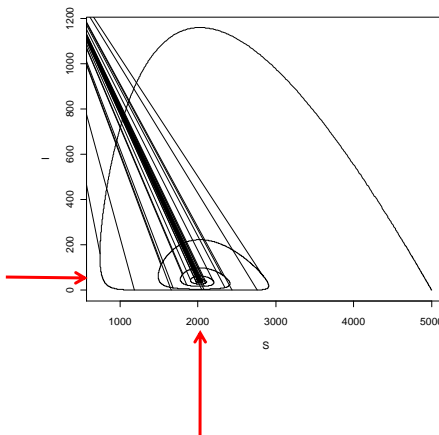
The Inter epidemic period



Equilibrium



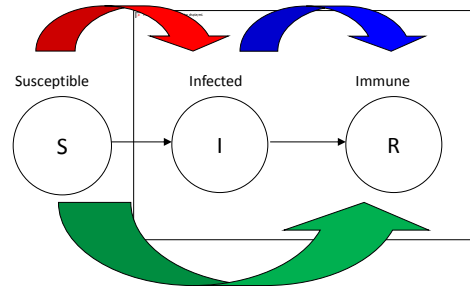
Equilibrium



At the long run, the infection reach the endemic equilibrium state in which at each time unit there are the same number of susceptible and infected individuals in the population.

Transmission Model with vaccination

$$\begin{aligned}\frac{dS(t)}{dt} &= N\mu \times (1-P) - \beta IS - \mu S \\ \frac{dI(t)}{dt} &= \beta IS - (\sigma + \mu)I \\ \frac{dR(t)}{dt} &= \sigma I - \mu R + N\mu \times P\end{aligned}$$



```

SIR<-function(t,state,parameters)
{
  with(as.list(c(state, parameters)),
  {
    dX <- 5000*mu*(1-P)-beta*Y*X - mu*X
    dY <- beta*Y*X - v*Y - mu*Y
    dZ <- v*Y - mu*Z+5000*mu*P
    list(c(dX, dY, dZ))
  })
}
  
```

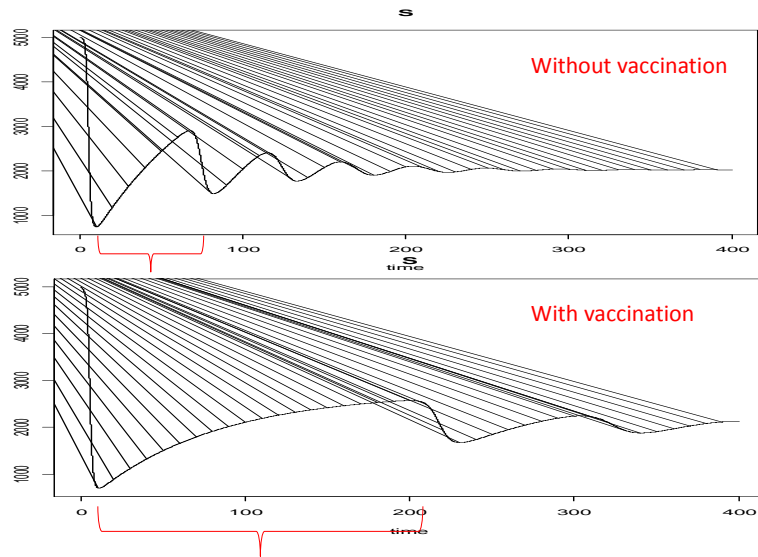
SIR model with vaccination at birth

45% are vaccinated at birth.

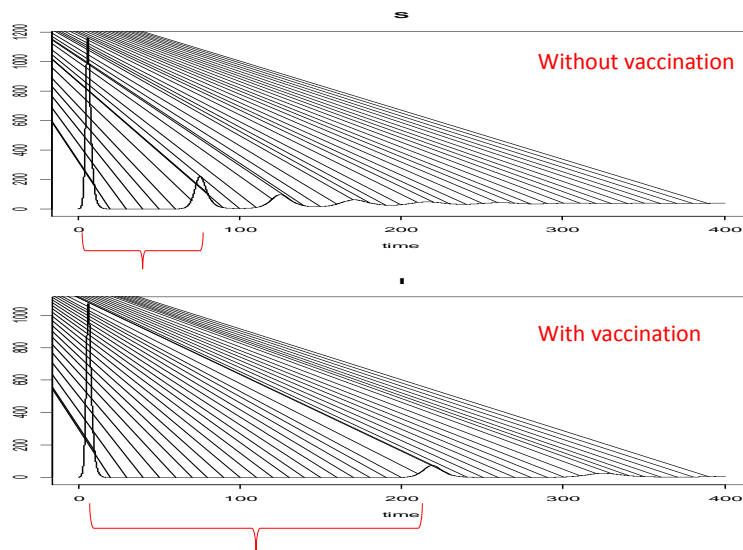
```

> parameters <- c(mu=1/75,beta=0.001/2, v=1, P=0.45)
> parameters
      mu      beta      v      P
0.01333333 0.00050000 1.00000000 0.45000000
  
```

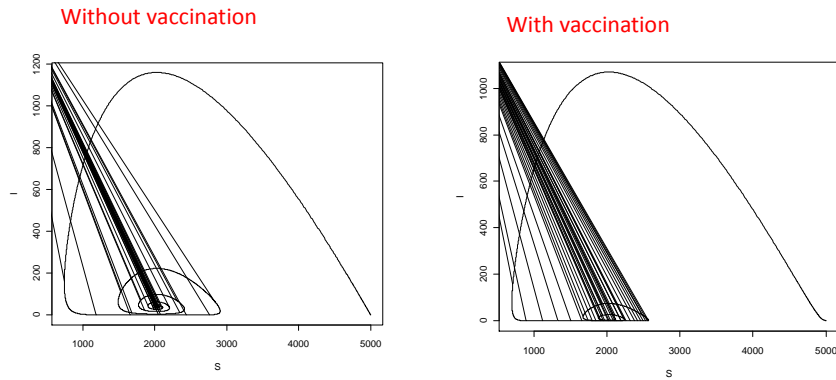
Susceptible: inter epidemic period



Infected class: inter epidemic period



Equilibrium



Equilibrium

Without vaccination

```
> out[39998:40001,]
      time      X      Y      Z
39998 399.97 2024.816 39.54316 2935.641
39999 399.98 2024.812 39.54280 2935.645
40000 399.99 2024.809 39.54243 2935.649
40001 400.00 2024.805 39.54206 2935.653
```

With vaccination

```
> out[39998:40001,]
      time      X      Y      Z
39998 399.97 2117.341 9.663568 2872.995
39999 399.98 2117.323 9.667949 2873.009
40000 399.99 2117.305 9.672332 2873.023
40001 400.00 2117.287 9.676716 2873.036
```