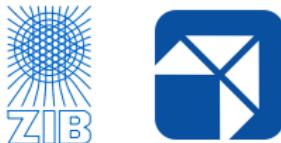


# Parameter estimation of a tuberculosis model in a patchy environment in Cameroon

D. P. Moualeu

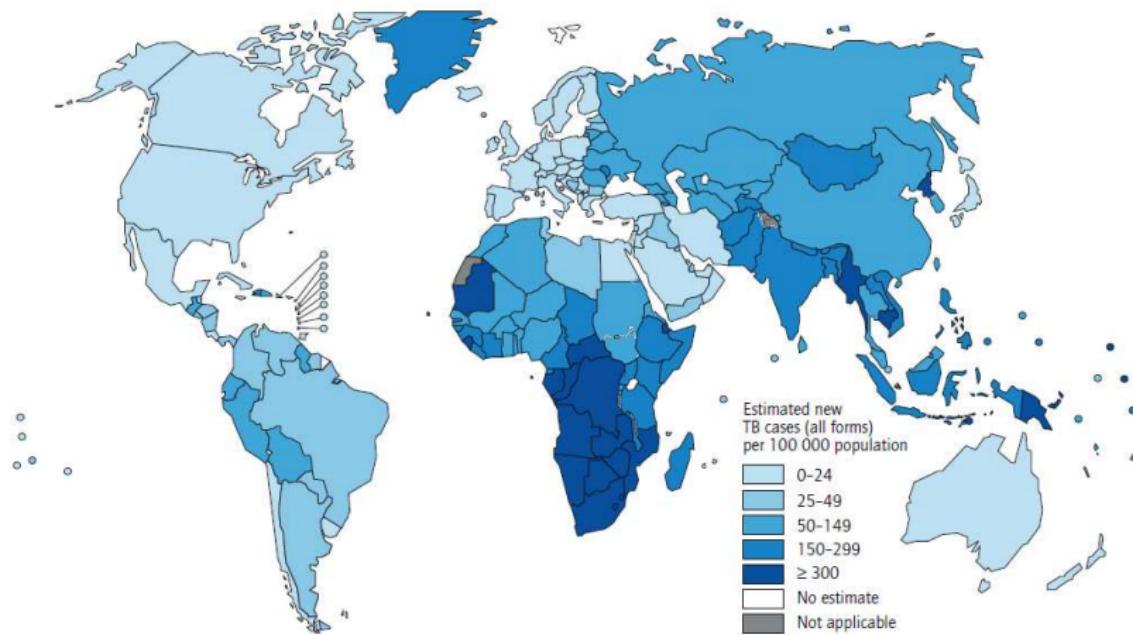
Joint work with S. Bowong, J. Kurths, P. Deuflhard



Biomat Conference 2013

Toronto, November 4, 2013

# TB distribution 2012



[WHO, 2012]

# Tuberculosis in Cameroon



## TB Notification 2011

- ▶ New pulmonary TB Cases: 19868
- ▶ Men 15-54 years old: 40%
- ▶ Women 15-54 years old: 51.54%
- ▶ HIV test positive: 38.91%
- ▶ Newly enrolled in HIV cases: 36458

## TB Outcome 2010

- ▶ New pulmonary smear-positive cases cured: 63.97%
- ▶ New pulmonary smear-positive cases died: 5.66%
- ▶ Quitting treatment: 9.16%

## TB Estimations in 2011 (for 100,000 population)

- ▶ Estimated incidence of TB: 243
- ▶ Estimated HIV in TB incidence: 76
- ▶ Estimated number of deaths from TB : 24
- ▶ Estimated Prevalence of deaths from TB : 299

Source: [\[vidiani.com/maps\]](http://vidiani.com/maps)

Source: [\[who.int/tb/country/en/\]](http://who.int/tb/country/en/)

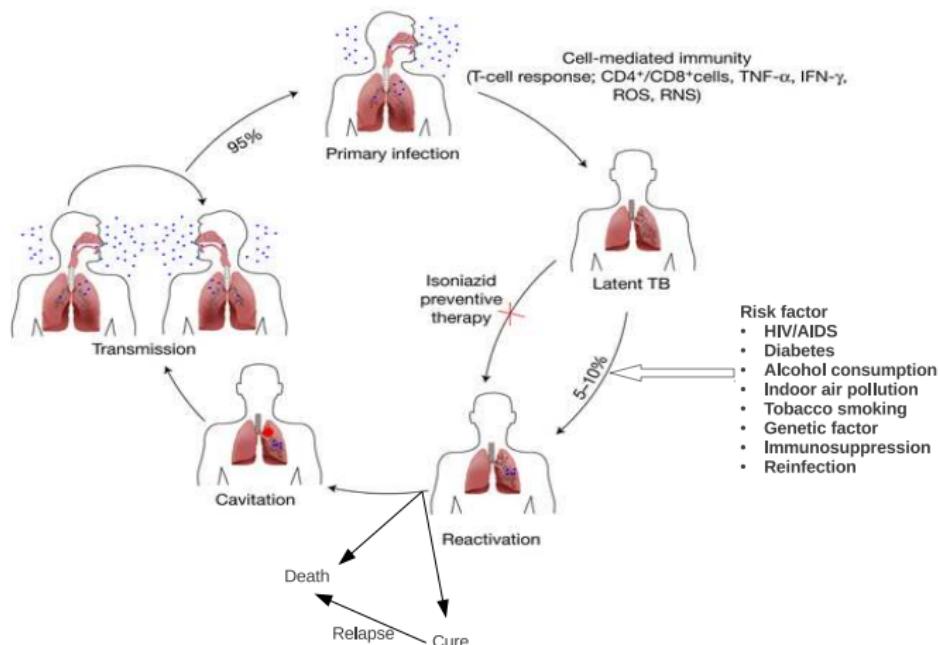
## Aims

- ▶ A model for transmission dynamics of TB among Cameroon's regions
- ▶ Parameter identification of the model

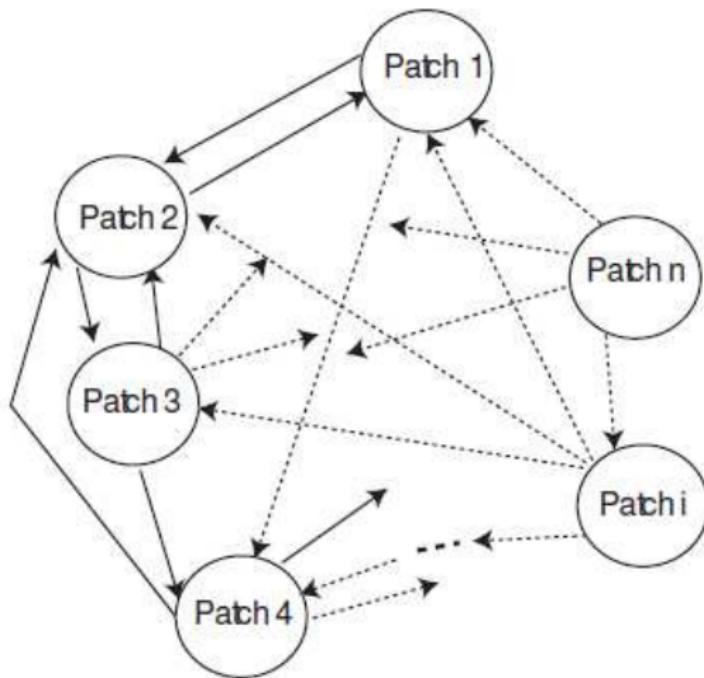
## What can we use such model for?

- ▶ Explore the role of undiagnosed infectious on TB transmission
- ▶ Identify key parameters on TB dynamics
- ▶ Determine sensitivities to changes in parameter values
- ▶ Estimate key parameters from measurable data

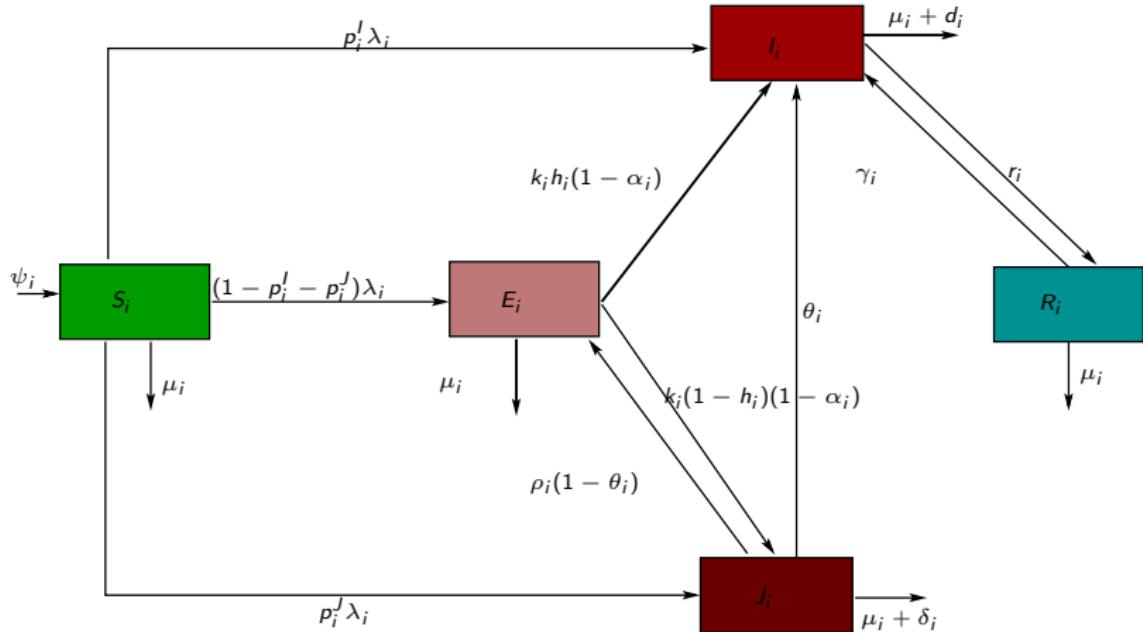
# TB life cycle



# Migration model



# Components and compartments



Force of infection:  $\lambda_i = \beta_i^I \frac{I_i}{N_i} + \beta_i^J J_i,$

## Model

$$\begin{cases} \dot{S} = \psi - \lambda \cdot S - \mu S + \mathcal{M}S \\ \dot{E} = (\mathbf{1} - p^I - p^J) \cdot \lambda \cdot S + \rho \cdot (\mathbf{1} - \theta) J - A_E \cdot E + \mathcal{M}E \\ \dot{I} = p^I \cdot \lambda \cdot S + \theta \cdot J + \gamma \cdot R + h \cdot (\mathbf{1} - \alpha) \cdot k \cdot E - A_I \cdot I + \eta \mathcal{M}I \\ \dot{J} = p^J \cdot \lambda \cdot S + (\mathbf{1} - h) \cdot (\mathbf{1} - \alpha) \cdot k \cdot E - A_J \cdot J + \eta \mathcal{M}J \\ \dot{R} = r \cdot I - A_R \cdot R + \mathcal{M}R \end{cases}$$

# Basic reproduction ratio

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$$\begin{cases} \dot{x} = \varphi(x) + \mathcal{M}S - \lambda(I)x, \\ \dot{y} = \mathcal{F}(x, y) - \mathcal{V}(x, y), \end{cases}$$

where  $x = S \in \mathbb{R}^n$  and  $y = (E, I, J, R) \in \mathbb{R}^{4n}$ .

$$\mathcal{R}_0 = \rho(\mathcal{F}_y(x_0, 0) \cdot (\mathcal{V}_y(x_0, 0))^{-1})$$

- ▶ Basic reproduction ratio ( $\mathcal{R}_0$ ): number of active cases one infectious generates on average over the course of its infectious period
- ▶  $\mathcal{R}_0 \leq 1$ : TB dies out in the long run
- ▶  $\mathcal{R}_0 > 1$ : TB persists and spread

- ▶ Model

$$\frac{d}{dt}y(t, \mathbf{p}) = f(t, y, \mathbf{p}), \quad t \geq 0$$

- ▶ Initial states vector  $y(0, \mathbf{p}) = y_0 \in \mathbb{R}^N$
- ▶ Vector of parameters  $\mathbf{p} \in \mathbb{R}^q$

$$\mathbf{S}_{ij}(t) := \left( \frac{\partial y_i}{\partial \mathbf{p}_j}(t) \right), \quad i = 1, \dots, N, \quad j = 1, \dots, q$$
$$\mathbf{S}' = f_y(y, \mathbf{p})\mathbf{S} + f_{\mathbf{p}}(y, \mathbf{p}), \quad \mathbf{S}(t_0) = 0$$

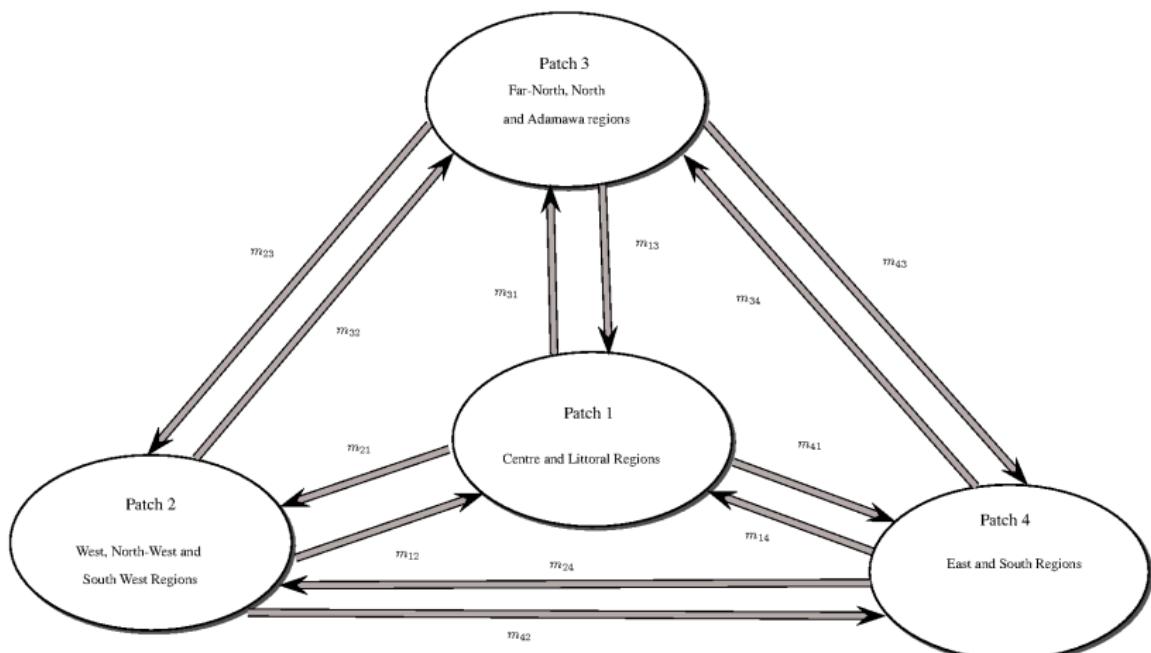
- ▶ Analysis of  $\mathbf{S}$  :

$$\mathbf{S}_j(t_k) := (\mathbf{S}_{1j}(t_k), \dots, \mathbf{S}_{nj}(t_k))^T$$

$$\mathbf{S}_{ij}(t_k) = \left( \frac{\partial y_{k_i}(t_k, \mathbf{p})}{\partial \mathbf{p}_j} \right), \quad k_i = 1, \dots, N$$

- ▶ Small sensitivity norms  $\|\mathbf{S}_j\| = \|(\mathbf{S}_{ij}(t))\|$ ,  $i = 1, \dots, m$  indicate a small sensitivity

# Application on 4 patches



# Sensitivity analysis (results)

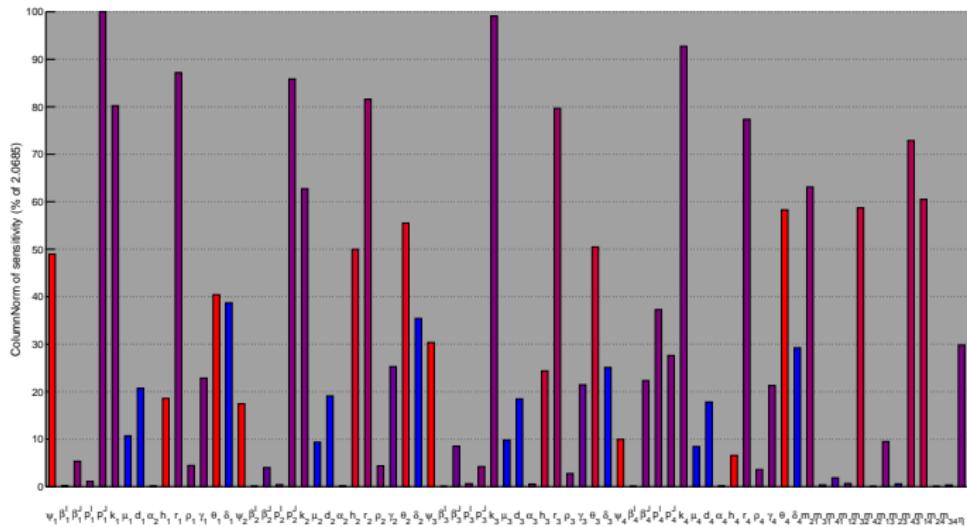
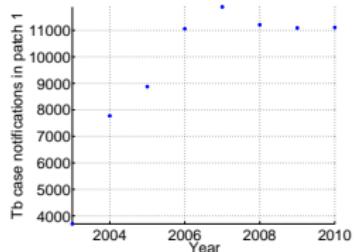
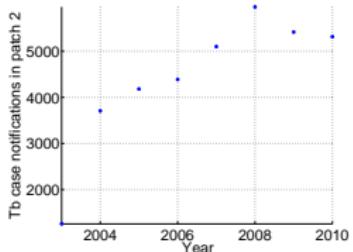


Figure: Sensitivity norms of unknown parameters

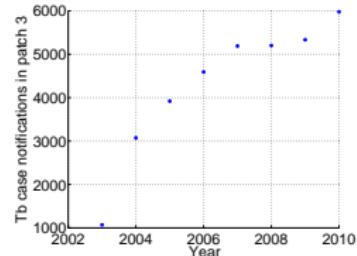
# Data: number of diagnosed infectious



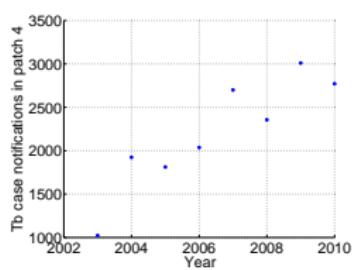
(a)  $I_1$



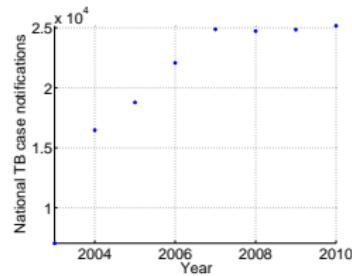
(b)  $I_2$



(c)  $I_3$

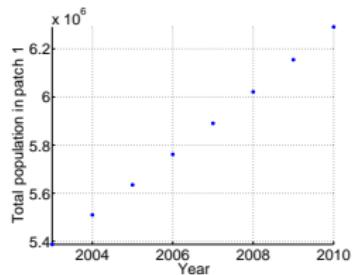


(d)  $I_4$

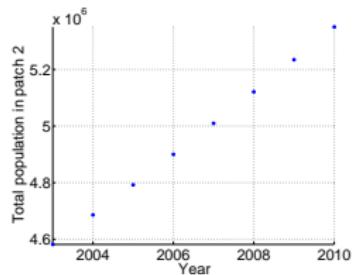


(e)  $I$

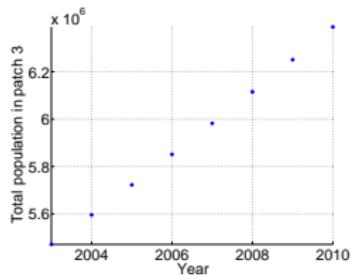
# Data: total population



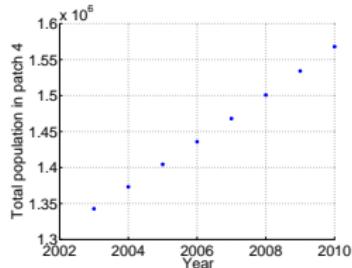
(f)  $N_1$



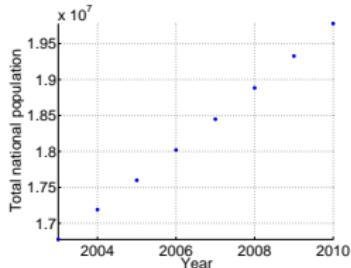
(g)  $N_2$



(h)  $N_3$



(i)  $N_4$



(j)  $N$

- ▶ Least squares formulation

$$\|F(\mathbf{p})\|_2^2 = \frac{1}{m} \sum_{j=1}^m \left\| \frac{y(\tau_j, \mathbf{p}) - z_j}{\delta z_j} \right\|_2^2 \longrightarrow \min$$

- ▶ Data  $z = (z_1, \dots, z_m)$
- ▶  $\delta z_i$  : measurement accuracy of  $z_i$
- ▶ The relative measurement accuracy is  $\delta z_i = \varepsilon_z z_i$  with  $\varepsilon_z = 10^{-1}$  to  $10^{-3}$  in experiments

- ▶ Solution of the **nonlinear** least squares problem by a global adaptive Gauss-Newton method

$$\begin{aligned} \| F'(\mathbf{p}^k) \cdot \Delta \mathbf{p}^k + F(\mathbf{p}^k) \|_2^2 &\rightarrow \min, \\ \mathbf{p}^{k+1} &= \mathbf{p}^k + \lambda_k \Delta \mathbf{p}^k, \quad k = 0, 1, 2, \dots \end{aligned}$$

[P. Deuflhard: Newton Methods for Nonlinear Problems, 2004]

- ▶ Sequence of **linear** least squares problems with  $(m \times q)$  Jacobian matrix  $F'(\mathbf{p})$
- ▶ A closer look of the expression of  $F'(\mathbf{p})$  reveals that

$$F'(p) = \left( \frac{1}{\delta z_j} \right) \mathcal{S}, \quad \mathcal{S} = (\mathbf{S}(t_1), \dots, \mathbf{S}(t_m))$$

- ▶ **Good initial guess** required

- ▶ QR factorization with column pivoting

$$F'(\mathbf{p}^k)\Pi = QR, r_{11} \geq r_{22} \geq \cdots \geq r_{qq}$$

[P. Deuflhard and A. Hohmann, 2003]

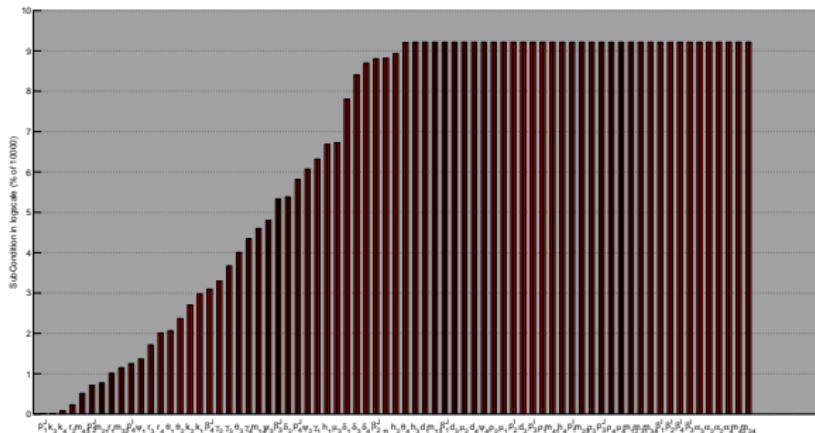
- ▶ The sub-condition of parameter  $\mathbf{p}_j$  is defined by

$$sc_j = \frac{r_{11}}{r_{jj}}$$

- ▶ Identifiable parameters:  $\varepsilon_{\mathbf{p}_j} sc_j < 1$

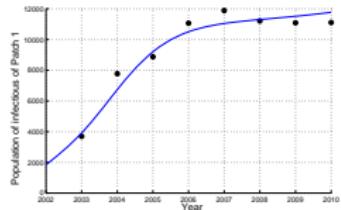
[P. Deuflhard and Sautter, 1980]

# Identifiability (results)

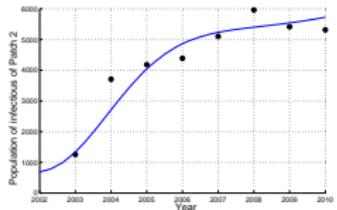


- ▶ Dependencies between parameters
- ▶ Only a few parameters can be identified

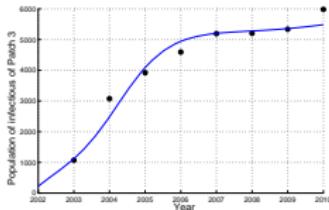
# Numerical results (I)



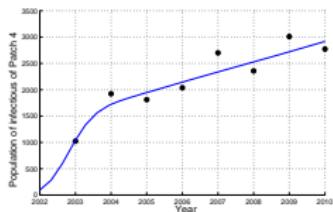
(k)  $I_1$



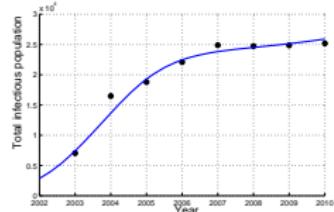
(l)  $I_2$



(m)  $I_3$

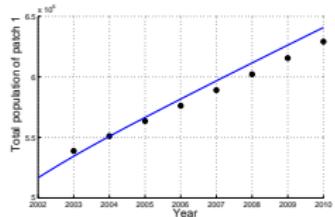


(n)  $I_4$

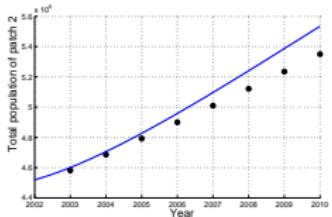


(o)  $I$

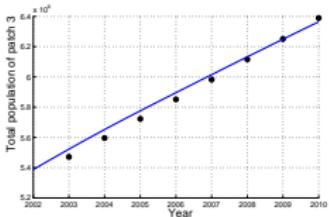
# Numerical results (N)



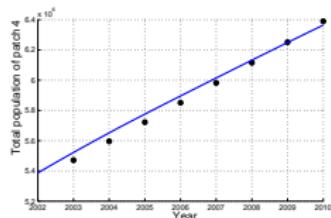
(p)  $N_1$



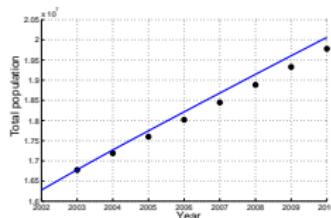
(q)  $N_2$



(r)  $N_3$

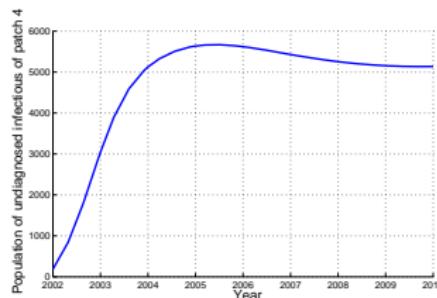
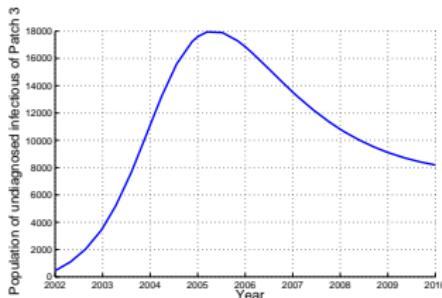
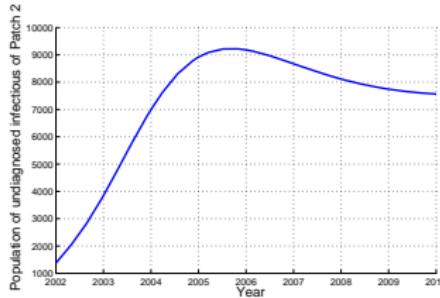
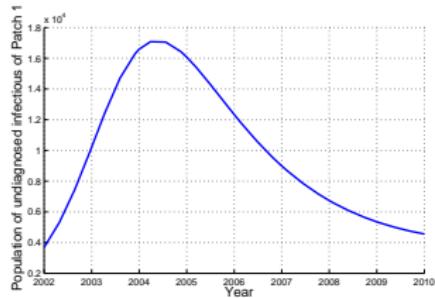


(s)  $N_4$

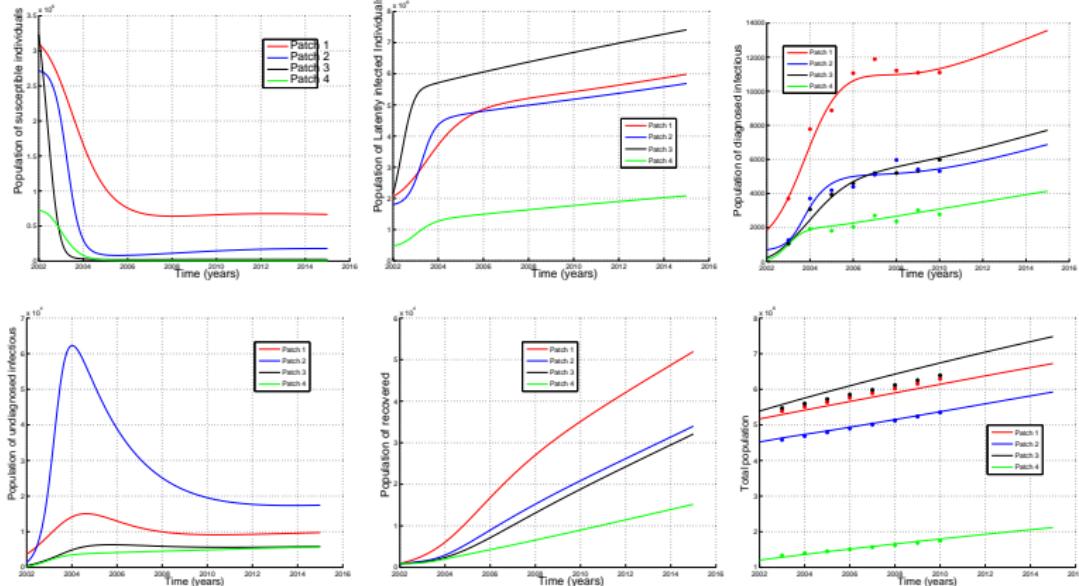


(t)  $N$

# Numerical results (J)



# Numerical results (J)



# Acknowledgements

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