How host heterogeneity governs tuberculosis reinfection?

M. Gabriela M. Gomes 1,* , Ricardo Águas 1,2 , João S. Lopes $^{1},$ Marta C. Nunes^{1,3}, Carlota Rebelo⁴, Paula Rodrigues^{1,5} and Claudio I. Struchiner⁶

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Instituto Gulbenkian de Ciência, Apartado 14, 2781-901 Oeiras, Portugal
² Department of Infectious Disease Epidemiology, MPC Centre for Qutbreek Anglysis an 2 Department of Infectious Disease Epidemiology, MRC Centre for Outbreak Analysis and Modelling,

Imperial College London, London W2 1PG, UK
³Respiratory and Meningeal Pathogens Research Unit, University of the Witwatersrand, Johannesburg,

South Africa
⁴Faculdade de Ciências de Lisboa, Centro de Matemática e Aplicações Fundamentais, Campo Grande,

Edificio C6, 1749-016 Lisboa, Portugal
⁵Departamento de Matemática and Centro de Matemática e Aplicações, FCT UNL,

Quinta da Torre, 2829-516 Caparica, Portugal
⁶ Program for Scientific Computing, Oswaldo Cruz Foundation, Avenue Brasil 4365, 21045-900 Rio de

Janeiro, Brazil

Recurrent episodes of tuberculosis (TB) can be due to relapse of latent infection or exogenous reinfection, and discrimination is crucial for control planning. Molecular genotyping of Mycobacterium tuberculosis isolates offers concrete opportunities to measure the relative contribution of reinfection in recurrent disease. Here, a mathematical model of TB transmission is fitted to data from 14 molecular epidemiology studies, enabling the estimation of relevant epidemiological parameters. Meta-analysis reveals that rates of reinfection after successful treatment are higher than rates of new TB, raising an important question about the underlying mechanism. We formulate two alternative mechanisms within our model framework: (i) infection increases susceptibility to reinfection or (ii) infection affects individuals differentially, thereby recruiting high-risk individuals to the group at risk for reinfection. The second mechanism is better supported by the fittings to the data, suggesting that reinfection rates are inflated through a population phenomenon that occurs in the presence of heterogeneity in individual risk of infection. As a result, rates of reinfection are higher when measured at the population level even though they might be lower at the individual level. Finally, differential host recruitment is modulated by transmission intensity, being less pronounced when incidence is high.

Keywords: tuberculosis epidemiology; transmission dynamics; partial immunity; reinfection

1. INTRODUCTION

Despite significant improvements in tuberculosis (TB) treatment over recent years, adequately treated patients are still at high risk of developing recurrent pulmonary disease (defined as an episode of TB following the cure of a previous episode). Recent estimates for the recurrence rate of TB across different regions point to an average of 2290 cases per 100 000 person-years at 12 months after treatment completion. In high-incidence regions, the average TB recurrence rate can reach 7850 per 100 000 person-years [[1](#page-4-0)].

The contribution of exogenous reinfection with Mycobacterium tuberculosis (Mtb) versus that of endogenous reactivation (relapse) of latent Mtb to the overall rate of recurrence of pulmonary disease is subject to controversies because these two mechanisms cannot be easily

* Author for correspondence (ggomes@igc.gulbenkian.pt).

disentangled. Deciphering the weight of each of these mechanisms is of great importance for policy-making. Advances in DNA fingerprinting techniques allowed the genotyping of the Mtb, causing different disease episodes [\[2](#page-4-0)]. These methods can reveal whether a new episode of disease is caused by infection with the same strain that caused a previous episode or a different one, enabling a classification into relapse or reinfection, respectively.

Longitudinal data from molecular epidemiological studies on TB reinfection have shown a positive correlation between the proportion of reinfection in recurrent cases and local incidence [\[3\]](#page-4-0). A long-term study, in an area of South Africa, with particularly high incidence, attributed the majority (77%) of recurrent TB cases to reinfection [\[4](#page-4-0)]. Moreover, the rate of TB reinfection was found to be four times higher than that of new TB, raising an important question about the underlying mechanism. Two possibilities have been proposed to explain these results: (i) infection increases susceptibility to reinfection or (ii) infection occurs at a higher rate in a high-risk subpopulation [[5](#page-4-0)].

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Figure 1. Diagram illustrating host selection. In a population with a heterogeneous susceptibility profile, infection tends to affect individuals at higher risk. This skews the distribution of recovered individuals towards higher susceptibility, which explains the increased reinfection rates in this subpopulation. Host selection is modulated by transmission intensity and its effect is weaker under higher transmission.

We address this question by constructing a general framework for TB transmission that encapsulates both mechanisms, and assess its capability to fit available data relating TB incidence and reinfection proportion among recurrent cases, in 14 regions throughout the world. The dataset was gathered by systematic literature review. The model postulates that some individuals are a priori more likely to be infected owing to enhanced susceptibility or exposure [[6](#page-4-0)–[8\]](#page-4-0). Infection is more likely to affect individuals at higher risk, who will naturally be over-represented in the treated subpopulation to a degree that is modulated by transmission intensity (illustrated in figure 1). This process alone acts to inflate the rate of reinfection at the population level, even if infection confers partial protection at the individual level. The relative susceptibility of individuals who have been previously infected over those who are naive is represented by a parameter σ , the value of which will be estimated by the fitting procedure.

2. METHODS

(a) Literature review

Through a systematic literature review, we aggregated data on recurrent TB and its relationship with TB incidence. Published epidemiological studies were located via PubMed through searches on the following terms: tuberculosis, recurrent, relapse, reinfection or re-infection. Studies were included in the analysis if fulfilling the following criteria:

- study reports the number of recurrent TB cases, defined positive culture after bacteriologically confirmed cure or complete treatment following a first episode;
- study reports more than 10 recurrent TB cases;
- study discriminates between reinfection and relapse by comparing Mtb DNA fingerprinting profiles of the initial and recurrent episodes; and
- population-based study published up to March 2011.

We extracted the data for the local incidence from the study papers whenever possible. When no incidence was reported in the study itself, we used the estimates for the year 2000 in the respective country, provided by the World Health Organization [[9\]](#page-4-0). The proportion of reinfection was defined as the ratio between the number of patients with reinfection and all recurrent TB cases. One study performed in Cape Town also provided the ratio between the rate of reinfection TB in successfully treated patients and the rate of new cases of TB [\[4](#page-4-0)].

(b) Mathematical formulation

A mathematical model [\[10](#page-4-0),[11\]](#page-4-0) is extended by enabling the risk of infection to be heterogeneously distributed among the population. The transmission model is replicated in two subpopulations, indexed by $i = 1,2$, such that immunologically naive individuals in subpopulation 1 (low risk) are subject to a per capita rate of infection $(\lambda_1 = \alpha_1 \lambda)$, whereas in subpopulation 2 (high risk) infection occurs at a higher rate ($\lambda_2 = \alpha_2 \lambda$), where $\alpha_1 < \alpha_2$, and γ_1 , γ_2 are the proportions of the population in each risk group. Within each group, individuals are classified—according to their infection history into susceptible (S_i) , primary infection (P_i) , latent (L_i) , active pulmonary tuberculosis (I_i) and recovered (R_i) . Latent and recovered individuals can be reinfected at a rate that is proportional to the rate of first infection, with multiplicative factor σ . The model is written as a system of differential equations

$$
\begin{aligned}\n\frac{dS_i}{dt} &= \mu \gamma_i - (\lambda_i + \mu) S_i \\
\frac{dP_i}{dt} &= \lambda_i S_i + \sigma \lambda_i (L_i + R_i) - (\delta + \mu) P_i \\
\frac{dI_i}{dt} &= \phi \delta P_i + \omega (L_i + R_i) - (\tau + \mu) I_i \\
\frac{dL_i}{dt} &= (1 - \phi) \delta P_i - (\sigma \lambda_i + \omega + \mu) L_i \\
\frac{dR_i}{dt} &= \tau I_i - (\sigma \lambda_i + \omega + \mu) R_i,\n\end{aligned}
$$
\n(2.1)

where μ (=1/70 yr⁻¹) is the birth and death rate, ϕ (=0.05) is the proportion of primary infections progressing to active pulmonary disease, and τ (=2 yr⁻¹) is the rate at which infectious individuals are detected and treated. An auxiliary parameter δ (=12 yr⁻¹) is included to represent the rate of progression from primary infection, although this can be interpreted only in conjugation with other parameters. For example, the rate of progression from primary infection to disease is $\phi\delta$ (=0.6 yr⁻¹). The value for μ is consistent with human populations with life expectancy of 70 years, whereas ϕ and τ are consistent with the medical literature [[12\]](#page-4-0). The rate of relapse (ω) is fixed at the higher end of published estimates ($\omega = 0.01 \text{ yr}^{-1}$) [\[13](#page-4-0)–[17](#page-4-0)], with justification and sensitivity analysis provided in the electronic supplementary material. Parameters referring to the reinfection factor (σ) and risk heterogeneity are estimated in order to adjust the equilibrium model solutions to the available data. By noting that γ_1 and γ_2 are proportions, and thus $\gamma_1 + \gamma_2 = 1$, and by normalizing the average risk factor such that $\gamma_1\alpha_1 + \gamma_2\alpha_2 = 1$, heterogeneity is fully parametrized by the low-risk parameters, γ_1 ($\equiv \gamma$) and α_1 ($\equiv \alpha$). The parameters are listed in [table 1](#page-2-0).

(c) Measures of incidence

TB incidence (cases per 100 000 person-years) is calculated from the equilibrium proportion of infectious individuals as

$$
Y = \left[\phi \delta \sum_{i=1}^{2} P_i + \omega \sum_{i=1}^{2} (L_i + R_i) \right] \times 100\,000
$$

= $(\tau + \mu) \sum_{i=1}^{2} I_i \times 100\,000.$ (2.2)

Table 2. Data collected from the systematic literature review.

Table 1. Model parameters.

	symbol definition	value
μ	birth and death rate	$1/70 \text{ yr}^{-1}$
	proportion progressing from primary infection to disease	0.05
	rate of progression from primary infection	12 yr^{-1}
	rate at which infectious individuals are detected and treated	2 yr^{-1}
ω	rate of endogenous reactivation	$0.01 \,\mathrm{yr}^{-1}$
σ	reinfection factor	estimated
	proportion low-risk group	estimated
α	low-risk factor	estimated

Following the criteria used in the data collection, we classify a recurrent TB case as an individual who enters the infectious class after having gone through the recovered class. This combines two pathways: relapse while in the recovered class, Y_{relapse} ; and exogenous reinfection with progression to active disease, $Y_{\text{reinfection}}$ (direct or following a latent period). We derive these quantities formally from equation (2.1). The instantaneous rate of relapse after successful treatment is the sum of $R_i \rightarrow I_i$ transitions

$$
Y_{\text{relapse}} = \omega \sum_{i=1}^{2} R_i,\tag{2.3}
$$

while the instantaneous rate of reinfection after successful treatment is the sum of $R_i \rightarrow P_i (\rightarrow L_i \rightarrow P_i)_n \rightarrow I_i$ and $R_i \to P_i(\to L_i \to P_i)_n \to L_i \to I_i$ transitions for any number (*n*) of iterations of the cycle $P_i \rightarrow L_i \rightarrow P_i$, derived as

$$
Y_{\text{reinfection}} = \sum_{i=1}^{2} \left\{ \sigma \lambda_i R_i \left[\frac{\phi \delta}{\delta + \mu} + \frac{(1 - \phi) \delta \omega}{(\delta + \mu)(\omega + \mu + \sigma \lambda_i)} \right] \right\}
$$

$$
\times \sum_{n=0}^{\infty} \left[\frac{\sigma \lambda_i (1 - \phi) \delta}{(\delta + \mu)(\omega + \mu + \sigma \lambda_i)} \right]^n \right\}
$$

$$
= \sum_{i=1}^{2} \left\{ \sigma \lambda_i R_i \left[\frac{\phi \delta}{\delta + \mu} + \frac{(1 - \phi) \delta \omega}{(\delta + \mu)(\omega + \mu + \sigma \lambda_i)} \right] \right\}
$$

$$
\times \frac{(\delta + \mu)(\omega + \mu + \sigma \lambda_i)}{(\delta + \mu)(\omega + \mu) + \sigma \lambda_i(\mu + \phi \delta)} \right\}.
$$
(2.4)

Hence, the proportion of reinfection over all recurrent TB after successful treatment is

$$
p = \frac{Y_{\text{reinfection}}}{Y_{\text{relines}} + Y_{\text{reinfection}}}.\tag{2.5}
$$

The rate of new TB is given by

$$
Y_{\text{new}} = \sum_{i=1}^{2} \left\{ \lambda_i S_i \left[\frac{\phi \delta}{\delta + \mu} + \frac{(1 - \phi) \delta \omega}{(\delta + \mu)(\omega + \mu + \sigma \lambda_i)} \right] \times \frac{(\delta + \mu)(\omega + \mu + \sigma \lambda_i)}{(\delta + \mu)(\omega + \mu) + \sigma \lambda_i(\mu + \phi \delta)} \right\}.
$$
\n(2.6)

Finally, for comparison with the Cape Town study [[4\]](#page-4-0), we define κ as the ratio of the rate of reinfection TB among successfully treated patients over the rate of new TB cases among those who have never had a TB episode,

a WHO estimates for TB incidence in 2000.

formally calculated as

$$
\kappa = \frac{Y_{\text{reinfection}}/(R_1 + R_2)}{Y_{\text{new}}/(S_1 + S_2)}.\tag{2.7}
$$

(d) Meta-analysis

We perform a meta-analysis of the relationship between the proportion of reinfection in recurrent TB (p) and the incidence of TB (Y) by fitting the model described in equation (2.1) to the dataset collected by the systematic literature review. By taking the incidence rate as an independent variable of a nonlinear regression, we estimate the set of model parameters that best describe the observed trends in the proportion of TB reinfection. First, we have considered a model where the host population is homogeneous with respect to risk of infection, and proceeded to assess whether heterogeneity would significantly improve the ability of the model to fit to the data. A Gaussian–Newton algorithm was implemented to fit the model output to the data according to the least-squares criterion. We used an F-test and a log-likelihood test to assess whether the heterogeneous model provides a significantly better fit to the data.

3. RESULTS

We conducted a systematic literature review and found 14 molecular epidemiological studies reporting the proportion of reinfection in recurrent TB across communities presenting a wide range of endemic levels [\[3,4](#page-4-0),[18](#page-4-0)–[29\]](#page-5-0). Table 2 summarizes the data collected.

Figure 2. Proportion (p) of reinfection among recurrent tuberculosis (TB) cases as a function of TB incidence. Open circles correspond to the data described in [table 2](#page-2-0). Curves correspond to the best-fitting model under the two assumptions: homogeneous host population (dotted) and heterogeneous host population (dashed).

Table 3. Estimated parameter values.

parameter	heterogeneous model	homogeneous model
α σ	0.98 [0.95, 1.00] ^a 0.15 [0.00, 0.56] ^a 0.51 [0.00, 2.37] ^a	n.a. n.a. 3.87 $[1.61, 7.79]$ ^a
residual sum of squares	0.30	0.74
standard error estimate	0.16	0.24
F -test log-likelihood test	$8.12 (0.007)^{b}$ 12.70 $(0.002)^{b}$	n.a. n.a.

^aMode [95% CI]. b Score (p-value).

Figure 2 shows the output of the homogeneous and heterogeneous versions of equation (2.1) that best fit the dataset, whereas the estimated parameters are listed in table 3. The figure shows the proportion of reinfection TB in recurrent cases (p) versus TB incidence (Y) at equilibrium. The model shows a markedly nonlinear relationship between the proportion of reinfection and local incidence, which was not captured by previous studies [\[3](#page-4-0)]. The ratio (κ) of the rate of reinfection TB over the rate of new TB predicted by the model is plotted in figure 3, where it can be confronted with the measure obtained for Cape Town [[4](#page-4-0)]. Although the data published by the other 13 studies do not enable the calculation of this quantity, the homogeneous model predicts equally high values, whereas under the heterogeneous model, we expect higher ratios in all regions that report lower incidence than Cape Town.

Noting that the homogeneous model is nested within the heterogeneous, we have calculated the F -test statistic

Figure 3. Model predictions for the ratio (κ) of the rate of reinfection tuberculosis (TB) after successful treatment by the rate of new TB as a function of TB incidence. Open circle corresponds to the Cape Town study [\[4](#page-4-0)]. Curves are generated by the best-fitting models identified in figure 2: homogeneous host population (dotted) and heterogeneous host population (dashed).

to show that heterogeneity enables a significantly better fit to the data (table 3). The best scenario provided by this analysis estimates the heterogeneity parameters as $\gamma = 0.98$ and $\alpha = 0.15$, suggesting that the risk of infection is about 40 times higher than average in the 2 per cent subpopulation at highest risk. Although these are necessarily crude approximations that enabled the model to simultaneously fit a vast spectrum of epidemiological scenarios, they can serve as a basis for further resolution in a region-specific manner. The estimated reinfection factor (σ = 0.51) indicates that previous infection has a protective effect, contrary to a model proposed previously [\[30\]](#page-5-0).

4. DISCUSSION

We propose a minimal model for TB transmission to describe the relative contributions of reinfection and relapse to recurrent TB across a range of transmission intensities. A nonlinear relation between the proportion of reinfection and the local incidence is derived by fitting this mechanistic model to the dataset resulting from a systematic literature review. By accounting for heterogeneity in the risk of infection, we obtain significantly better model fittings to epidemiological data. This trend is in agreement with current understanding of TB transmission, especially in regions of low to moderate transmission, where TB is confined to particular risk groups with sporadic small outbreaks in the general population [\[31,32\]](#page-5-0), and has been previously noted in theoretical studies [[33](#page-5-0)]. Infection acts upon this variation and predominantly recruits those individuals at higher risk to the recovered category, thus inflating the rate of reinfection (as illustrated in [figure 1\)](#page-1-0). As a result,

population measures of reinfection rates in relation to first infection (κ) are higher than the reinfection factor at the individual level, σ .

The model predicts that, under heterogeneity, regions of low to moderate transmission support relatively higher κ than regions of high transmission. This is again owing to the way differential recruitment acts upon individuals at higher risk. For highly endemic regions, transmission intensity tends to homogenize the distributions of both susceptible and recovered individuals, making differential recruitment less pronounced [\[34\]](#page-5-0). Cape Town is the only study reporting the information required to estimate this ratio. Similar studies providing this data for other regions would be very valuable to confirm the validity and increase the accuracy of the results reported here.

The parameter estimation procedure (provided in the electronic supplementary material) supports higher relapse rates than those previously stated for European countries [14,15], whereas recent studies in African [13] and Asian [16,17] regions suggest values that are compatible with those considered here. This may be due to higher prevalence of co-infection with HIV in those settings or simply reflect regional differences in nutrition, smoking patterns, environmental conditions, population structure or the natural history of TB [16,17,[35](#page-5-0),[36](#page-5-0)]. Despite these acknowledged differences, we have opted for constancy across regions in model parameters, with the exception of the force of infection, because our objective is to make inferences about the global epidemiology of TB. These inferences were enabled by further assuming regional equilibrium conditions and selecting the set of model parameter values that best reproduce the trends of reinfection rates in relation to TB incidence.

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