# Supplementary material: A unified stochastic modelling framework for the spread of nosocomial infections

Martín López-García<sup>1</sup><sup>\*</sup>, Theodore Kypraios<sup>2</sup>

<sup>1</sup> School of Mathematics, University of Leeds, LS2 9JT Leeds, UK <sup>2</sup> School of Mathematical Sciences, University of Nottingham, NG7 2RD Nottingham, UK

## Abstract

In this Supplementary Material, we discuss about how systems of equations in Ref. [1] can be represented in matrix form and iteratively solved. Moreover, we report here parameter values considered in case studies 1-5 in Ref. [1], obtained from Refs. [2, 3, 4, 5, 6], and summarise in Table S6 the function rates  $\lambda_j(i_1,\ldots,i_M)$ ,  $\mu_j(i_1,\ldots,i_M)$  and  $\delta(i_1,\ldots,i_M)$  for these case studies.

#### 1 Matrix-oriented solutions

<sup>2</sup> 1.1 Number of infections caused by an individual at compartment j until he/she is removed or the outbreak is detected

<sup>5</sup> The objective here is to compute probabilities

$$
\nu_{(i_1,\ldots,i_M)}^{(j)}(n) = \mathbb{P}(R_{(i_1,\ldots,i_M)}^{(j)} = n), \quad n \ge 0,
$$

by solving the systems of equations given by  $[1, Eq. (2.3)].$ 

We can rewrite this system into a matrix equation of the <sup>8</sup> form

$$
\mathbf{D}^{(j)}\boldsymbol{\nu}^{(j)}(n) = \mathbf{e}^{(j)}(n), \tag{1}
$$

where matrix  $\mathbf{D}^{(j)}$  is independent of the value  $n \geq 0$ , while column vectors  $v^{(j)}(n)$  and  $e^{(j)}(n)$  depend on this value. In particular,

$$
(\mathbf{D}^{(j)})_{(i_1,\ldots,i_M),(i_1,\ldots,i_k-1,\ldots,i_M)} = \frac{1}{\theta_{(i_1,\ldots,i_M)}} \mu_k (i_k 1_{k \neq j}
$$
  
+ $(i_k - 1)1_{k=j}), \quad 1 \leq k \leq M,$   

$$
(\mathbf{D}^{(j)})_{(i_1,\ldots,i_M),(i_1,\ldots,i_k+1,\ldots,i_M)} = \frac{1}{\theta_{(i_1,\ldots,i_M)}} \left(\lambda_k + \sum_{l=1,\ l \neq j}^M \lambda_{lk} i_l + (i_j - 1)\lambda_{jk}\right) (N_k - i_k), \quad 1 \leq k \leq M,
$$
  

$$
(\mathbf{e}^{(j)}(n))_{(i_1,\ldots,i_M)} = \frac{1}{\theta_{(i_1,\ldots,i_M)}} \left(1_{n>0} \sum_{k=1}^M (N_k - i_k) \lambda_{jk}\right)
$$
  

$$
\times \nu_{(i_1,\ldots,i_k+1,\ldots,i_M)}^{(j)} (n-1) + 1_{n=0} (\mu_j + \delta(i_1,\ldots,i_M))\right).
$$

Dimensionality of system in Eq. (1) is not  $\#\mathcal{C}$  = <sup>10</sup>  $\prod_{k=1}^{M} (N_k + 1)$ , but  $N_j \prod_{k=1, k \neq j}^{M} (N_k + 1)$ , since probabilities  $\nu_{(i)}^{(j)}$  $\psi_{(i_1,\ldots,i_M)}^{(j)}(n)$  are only defined for states  $(i_1,\ldots,i_M) \in \mathcal{C}$  with  $i_j > 0$ . Moreover, we are storing in Eq. (1) probabilities  $\nu_{(i)}^{(j)}$  $\binom{1}{i_1,\ldots,i_M}(n)$  in the column vector 13

$$
\boldsymbol{\nu}^{(j)}(n) = \begin{pmatrix} \nu^{(j)}_{(0,0,\ldots,0,0)}(n) \\ \nu^{(j)}_{(0,0,\ldots,0,1)}(n) \\ \nu^{(j)}_{(0,0,\ldots,0,2)}(n) \\ \vdots \\ \nu^{(j)}_{(0,0,\ldots,0,N_M)}(n) \\ \nu^{(j)}_{(0,0,\ldots,1,0)}(n) \\ \nu^{(j)}_{(0,0,\ldots,1,1)}(n) \\ \nu^{(j)}_{(0,0,\ldots,1,2)}(n) \\ \vdots \\ \nu^{(j)}_{(N_1,N_2,\ldots,N_{M-1},N_M)}(n) \end{pmatrix}
$$

,

so that each row in this matrix system represents a state  $_{14}$  $(i_1, \ldots, i_M) \in \mathcal{C}$ , with  $i_j > 0$ . Due to the lexicographic order followed above when ordering these states <sup>16</sup> by rows, each state  $(i_1, \ldots, i_M)$  with  $i_j > 0$  corresponds 17 to the  $\sum_{k=1}^{M} (1_{k\neq j} i_k + 1_{k=j} (i_j-1)) \prod_{p=k+1}^{M} (1_{p\neq j} (N_p+1) +$  $(1_{p=j}N_p)^{th}$  row (*i.e.*, equation) in Eq. (1). Finally, since 19 matrix  $\mathbf{D}^{(j)}$  is significantly sparse, for numerical results in 20  $[1, Section 3]$  we solve this system of linear equations by 21 using the *scipy.sparse.linalg* Python package. This involves  $_{22}$ solving Eq. (1) for  $n = 0$ , and then iteratively solving it for 23 values  $n \geq 1$  using probabilities  $\nu_{(i)}^{(j)}$  $\binom{(J)}{(i_1,\ldots,i_M)}(n-1)$ , which are 24 stored in column vector  $v^{(j)}(n-1)$  previously computed. 25

<sup>∗</sup>Author for correspondence (m.lopezgarcia@leeds.ac.uk)

<sup>26</sup> 1.2 Number of infections caused by an  $27$  individual at compartment j, among  $28$  individuals at compartment k, until  $\hbar$  he/she is removed or the outbreak is <sup>30</sup> detected

<sup>31</sup> The objective here is to compute probabilities

$$
\nu_{(i_1,\ldots,i_M)}^{(j)}(k;n) \quad = \quad \mathbb{P}(R_{(i_1,\ldots,i_M)}^{(j)}(k)=n), \quad n \ge 0,
$$

 $32$  by solving the systems of equations given by [1, Eq.  $(2.2)$ ]. <sup>33</sup> Again, we can construct and iteratively solve matrix sys-

<sup>34</sup> tems of the form

$$
{\bf D}^{(j)}(k){\boldsymbol \nu}^{(j)}(k;n) \;\; = \;\; {\bf e}^{(j)}(k;n),
$$

where

$$
(\mathbf{D}^{(j)}(k))_{(i_1,...,i_M),(i_1,...,i_l-1,...,i_M)} = \frac{1}{\theta_{(i_1,...,i_M)}} \mu_l (i_l 1_{l \neq j}
$$
  
+  $(i_l - 1)1_{l=j}), \quad 1 \leq l \leq M,$   

$$
(\mathbf{D}^{(j)}(k))_{(i_1,...,i_M),(i_1,...,i_l+1,...,i_M)} = \frac{1}{\theta_{(i_1,...,i_M)}} \left( 1_{l \neq k} \left( \lambda_l \right) \right)
$$
  
+ 
$$
\sum_{p=1}^{M} \lambda_{pl} i_p (N_l - i_l) + 1_{l=k} (\lambda_l + \sum_{p=1, p \neq j}^{M} \lambda_{pl} i_p
$$
  
+ 
$$
\lambda_{jl} (i_j - 1)) (N_l - i_l) , \quad 1 \leq l \leq M,
$$
  

$$
(\mathbf{e}^{(j)}(k;n))_{(i_1,...,i_M)} = \frac{1}{\theta_{(i_1,...,i_M)}} \left( 1_{n>0} (N_k - i_k) \lambda_{jk} \right)
$$
  

$$
\times \nu_{(i_1,...,i_k+1,...,i_M)}^{(j)} (k; n-1) + 1_{n=0} (\mu_j + \delta(i_1,...,i_M)) \right)
$$

and where probabilities  $\nu_{(i)}^{(j)}$ <sup>35</sup> and where probabilities  $\nu_{(i_1,...,i_M)}^{(j)}(k;n)$  are stored in the <sup>36</sup> column vectors  $\nu^{(j)}(k; n)$ , as in subsection 1.1 of this Sup-<sup>37</sup> plementary Material.

## <sup>38</sup> 2 Parameter values for case studies  $39 \hspace{1.5cm} 1\text{-}5$

 In Tables S1-S5, we report parameter values considered in case studies 1-5 in Ref. [1], directly obtained from Refs. [2, 3, 4, 5, 6]. In Table S6, we summarise the 43 functional forms of rates  $\lambda_j(i_1,\ldots,i_M), \mu_j(i_1,\ldots,i_M)$  and  $\delta(i_1,\ldots,i_M)$  for case studies 1-5, according to the cor- responding model assumptions and model parameters de-scribed in Ref. [1].

### <sup>47</sup> References

- $_{48}$  [1] López-García M, Kypraios T (2018) A unified stochas-<sup>49</sup> tic modelling framework for the spread of nosoco-<sup>50</sup> mial infections. Journal of the Royal Society Interface <sup>51</sup> 20180060. http://dx.doi.org/10.1098/rsif.2018.0060.
- $52$  [2] Artalejo, JR (2014) On the Markovian approach for <sup>53</sup> modeling the dynamics of nosocomial infections. Acta <sup>54</sup> Biotheoretica, 62: 15-34.

	Meaning	Value
$N_p$	Number of patients	20
$N_{HCW}$	Number of HCWs	3
$\mu$	Patient discharge rate	0.1
$\gamma$	Patient detection rate	0.1
$\mu^{\prime}$	HCW hand-washing rate	14
ß	HCW-to-patient colonization rate	
$\overline{\beta'}$	Patient-to-HCW contamination rate	គី
$\sigma$	Fraction of admitted patients colonized	0.01

Table S1: Parameter values from Artalejo (2014) [2], for the spread of MRSA in an hypothetical intensive care unit. Time units: *days*. Case study 1

	Meaning	RICU
$N_P$	Number of patients	7
$N_{HCW}$	Number of HCWs	14
$N_V$	Number of volunteers	$\mathfrak{D}$
$\varphi$	Fraction of admitted patients colonized	0.165
	Length of stay, non-colonized patients	7
$\frac{\frac{1}{\delta_U}}{\frac{1}{\delta_C}}$	Length of stay, colonized patients	13
$\eta$	Hygienic level, HCW-patient	0.46
$\xi$	Hygienic level, volunteer-patient	0.23
$\beta_{PH}$	Patient-HCW contact rate	0.72
$\beta_{PV}$	Patient-volunteer contact rate	0.20
$\gamma_H$	HCW hand-washing rate	24
$\gamma_V$	Volunteer hand-washing rate	12

Table S2: Parameter values from Wang et al. (2011) [3], for the spread of MRSA in the Respiratory Intensive Care Unit (RICU) at Beijing Tongren Hospital. Time units: days. Case study 2

	Meaning	Value
$N_p$	Number of patients	20
$N_{s}$	Number of HCWs	5
$N_e$	Number of surfaces	100
$\phi$	Fraction of admitted patients colonized	0.1
$\gamma$	Discharge rate, non-colonized patients	0.1
$\gamma'$	Discharge rate, colonized patients	0.05
$\mu$	Staff decontamination rate	24
$\kappa$	Surfaces decontamination rate	1
$\beta_{sp}$	Staff-to-patient colonization rate	0.3
$\beta_{se}$	Staff-to-surface contamination rate	$\overline{2}$
$\beta_{ps}$	Patient-to-staff contamination rate	$\overline{2}$
$\beta_{pe}$	Patient-to-surface contamination rate	$\overline{2}$
$\beta_{es}$	Surface-to-staff contamination rate	$\overline{2}$
$\beta_{ep}$	Surface-to-patient colonization rate	0.3

Table S3: Parameter values from Wolkewitz et al. (2008) [4], for an VRE outbreak in the onco-haematological unit at the University Medical Center Freiburg in Germany. Time units: days. Case study 3

- [3] Wang J, Wang L, Magal P, Wang Y, Zhuo J, Lu X, <sup>55</sup> Ruan S (2011) Modelling the transmission dynamics 56 of meticillin-resistant Staphylococcus aureus in Bei- <sup>57</sup> jing Tongren hospital. Journal of Hospital Infection, sa 79: 302-308. <sup>59</sup>
- [4] Wolkewitz M, Dettenkofer M, Bertz H, Schumacher 60

,

				Rate function	
CS	M	$\mu_i(i_1,\ldots,i_M) = \mu_i i_i$		$\lambda_j(i_1,\ldots,i_M)=(N_j-i_j)\left(\lambda_j+\sum\lambda_{kj}i_k\right)$	$\delta(i_1,\ldots,i_M)$
$\mathbf{1}$	2	$\mu_1 = (1 - \sigma)\mu, \mu_2 = \mu'$	$\lambda_1 = \sigma \mu$ , $\lambda_2 = 0$	$\lambda_{12} = \beta', \lambda_{21} = \beta$	$\delta(i_1, i_2) = \gamma i_1$
$\overline{2}$	$\mathcal{S}$	$\mu_1 = \delta_C(1-\varphi), \mu_2 = \gamma_H$	$\lambda_1 = \delta_U \varphi, \lambda_2 = 0$	$\lambda_{12} = \frac{1-\eta}{N_P} \beta_{PH}, \lambda_{13} = \frac{1-\xi}{N_P} \beta_{PV}$	$\delta(i_1, i_2, i_3) = 0$
		$\mu_3 = \gamma_V$	$\lambda_3=0$	$\lambda_{21} = \frac{1-\eta}{N_P} \beta_{PH}, \lambda_{23} = 0$	
				$\lambda_{31}=\frac{1-\xi}{N_P}\beta_{PV},\,\lambda_{32}=0$	
3		3 $\mu_1 = \gamma'(1-\phi), \mu_2 = \mu$	$\lambda_1 = \gamma \phi$ , $\lambda_2 = 0$	$\lambda_{12} = \frac{\beta_{ps}}{N_p}, \lambda_{13} = \frac{\beta_{pe}}{N_p}$	$\delta(i_1, i_2, i_3) = 0$
		$\mu_3 = \kappa$	$\lambda_3=0$	$\lambda_{21} = \frac{\beta_{sp}}{N}, \lambda_{23} = \frac{\beta_{se}}{N}$	
				$\lambda_{31} = \frac{\beta_{ep}}{N}, \, \lambda_{32} = \frac{\beta_{es}}{N_{s}}$	
$\overline{4}$	$\overline{4}$	$\mu_i = \nu(1 - p_C), \ 1 \leq j \leq 4 \mid \lambda_i = \nu p_C + \lambda, \ 1 \leq j \leq 4$		$\lambda_{ik} = \beta_{DR}, 1 \leq j \neq k \leq 4$	$\delta(i_1,\ldots,i_4)=0$
				$\lambda_{ij} = \beta_{SR}, 1 \leq j \leq 4$	
$5^{\circ}$	11	$\mu_j = \gamma, 1 \leq j \leq 4$	$\lambda_i = 0, 1 \leq j \leq 11$	$\lambda_{51} = \lambda_{15} = \lambda_{62} = \lambda_{26} = \beta_{AP1}$	$\delta(i_1, \ldots, i_{11}) = 0$
		$\mu_i = \mu, 5 \leq j \leq 11$		$\lambda_{73} = \lambda_{37} = \lambda_{84} = \lambda_{48} = \beta_{AP1}$	
				$\lambda_{91}=\lambda_{19}=\lambda_{92}=\lambda_{29}=\beta_{AP2}$	
				$\lambda_{10,3} = \lambda_{3,10} = \lambda_{10,4} = \lambda_{4,10} = \beta_{AP2}$	
				$\lambda_{11,1} = \lambda_{1,11} = \lambda_{11,2} = \lambda_{2,11} = \beta_{Peri}$	
				$\lambda_{11,3} = \lambda_{3,11} = \lambda_{11,4} = \lambda_{4,11} = \beta_{Peri}$	
				For others $(j,k)$ : $\lambda_{jk} = 0$	

Table S6: Functional forms for case studies 1-5 (CS 1-5)

	Meaning	Value
$N_p$	Number of patients	9
$\nu$	Discharge rate	0.1
$p_C$	Fraction of admitted patients colonized	0.01
$\beta_{DR}$	Cross-colonization rate, different rooms	0.0238
$\beta_{SR}$	Cross-colonization rate, same room	0.0366
	Spontaneous colonization rate	0.0037

Table S4: Parameter values from López-García (2016) [5], for an MRSA outbreak in an intensive care unit with four rooms. Parameter values  $\nu$  and  $p_C$  from Artalejo et al. (2014) [2]. Time units: days. Case study 4

	Meaning	Value
$\beta_{AP1}$	Patient-AP1 transmission rate	0.35
$\beta_{AP2}$	Patient-AP2 transmission rate	0.12
$\beta_{Peri}$	Patient-peripatetic transmission rate	0.07
$\mu$	Hand-washing rate for all HCWs	$1 - 24$
	Length of stay for all patients	10

Table S5: Parameter values from Temime et al. (2009) [6], for a bacterial outbreak in an hypothetical intensive-care unit. Time units: days. Case study 5

 M, Huebner J (2008) Environmental contamination as an important route for the transmission of the hospi- tal pathogen VRE: modelling and prediction of classical interventions. Infectious Diseases: Research and Treat-<sup>65</sup> ment, 1.

- <sup>66</sup> [5] López-García M (2016) Stochastic descriptors in an <sup>67</sup> SIR epidemic model for heterogeneous individuals in <sup>68</sup> small networks. Mathematical Biosciences, 271: 42-61.
- <sup>69</sup> [6] Temime L, Opatowski L, Pannet Y, Brun-Buisson C, 70 Boëlle PY, Guillemot D (2009) Peripatetic health-care

workers as potential superspreaders. Proceedings of the  $\frac{71}{71}$ National Academy of Sciences, 106: 18420-18425.