

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

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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 $\lambda_j(i_1, \dots, i_M)$, $\mu_j(i_1, \dots, i_M)$ and $\delta(i_1, \dots, i_M)$ for these case studies.

1 Matrix-oriented solutions

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

The objective here is to compute probabilities

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

by solving the systems of equations given by [1, Eq. (3)]. We can rewrite this system into a matrix equation of the form

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

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

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

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

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

Dimensionality of system in Eq. (1) is not $\#\mathcal{C} = \prod_{k=1}^M (N_k + 1)$, but $N_j \prod_{k=1, k \neq j}^M (N_k + 1)$, since probabilities $\nu_{(i_1, \dots, i_M)}^{(j)}(n)$ are only defined for states $(i_1, \dots, i_M) \in \mathcal{C}$

with $i_j > 0$. Moreover, we are storing in Eq. (1) probabilities $\nu_{(i_1, \dots, i_M)}^{(j)}(n)$ in the column vector

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

so that each row in this matrix system represents a state $(i_1, \dots, i_M) \in \mathcal{C}$, with $i_j > 0$. Due to the lexicographic order followed above when ordering these states by rows, each state (i_1, \dots, i_M) with $i_j > 0$ corresponds 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 matrix $\mathbf{D}^{(j)}$ is significantly sparse, for numerical results in [1, Section 3] we solve this system of linear equations by using the *scipy.sparse.linalg* Python package. This involves solving Eq. (1) for $n = 0$, and then iteratively solving it for values $n \geq 1$ using probabilities $\nu_{(i_1, \dots, i_M)}^{(j)}(n-1)$, which are stored in column vector $\boldsymbol{\nu}^{(j)}(n-1)$ previously computed.

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1.2 Number of infections caused by an individual at compartment j , among individuals at compartment k , until he/she is removed or the outbreak is detected

The objective here is to compute probabilities

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

by solving the systems of equations given by [1, Eq. (2)]. Again, we can construct and iteratively solve matrix systems of the form

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

where

$$\begin{aligned} (\mathbf{D}^{(j)}(k))_{(i_1, \dots, i_M), (i_1, \dots, i_{l-1}, \dots, i_M)} &= \frac{1}{\theta_{(i_1, \dots, i_M)}} \mu_l (i_l \mathbf{1}_{l \neq j} \\ &+ (i_l - 1) \mathbf{1}_{l=j}), \quad 1 \leq l \leq M, \\ (\mathbf{D}^{(j)}(k))_{(i_1, \dots, i_M), (i_1, \dots, i_{l+1}, \dots, i_M)} &= \frac{1}{\theta_{(i_1, \dots, i_M)}} \left(\mathbf{1}_{l \neq k} (\lambda_l \right. \\ &+ \sum_{p=1}^M \lambda_{pl} i_p) (N_l - i_l) + \mathbf{1}_{l=k} (\lambda_l + \sum_{p=1, p \neq j}^M \lambda_{pl} i_p \\ &+ \lambda_{jl} (i_j - 1)) (N_l - i_l) \Big), \quad 1 \leq l \leq M, \\ (\mathbf{e}^{(j)}(k; n))_{(i_1, \dots, i_M)} &= \frac{1}{\theta_{(i_1, \dots, i_M)}} \left(\mathbf{1}_{n > 0} (N_k - i_k) \lambda_{jk} \right. \\ &\times \nu_{(i_1, \dots, i_k+1, \dots, i_M)}^{(j)}(k; n-1) + \mathbf{1}_{n=0} (\mu_j + \delta(i_1, \dots, i_M)) \Big), \end{aligned}$$

and where probabilities $\nu_{(i_1, \dots, i_M)}^{(j)}(k; n)$ are stored in the column vectors $\boldsymbol{\nu}^{(j)}(k; n)$, as in subsection 1.1 of this Supplementary Material.

2 Parameter values for case studies 1-5

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 functional forms of rates $\lambda_j(i_1, \dots, i_M)$, $\mu_j(i_1, \dots, i_M)$ and $\delta(i_1, \dots, i_M)$ for case studies 1-5, according to the corresponding model assumptions and model parameters described in Ref. [1].

References

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	Meaning	Value
N_p	Number of patients	20
N_{HCW}	Number of HCWs	3
μ	Patient discharge rate	0.1
γ	Patient detection rate	0.1
μ'	HCW hand-washing rate	14
β	HCW-to-patient colonization rate	$\frac{1}{6}$
β'	Patient-to-HCW contamination rate	$\frac{1}{6}$
σ	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	2
φ	Fraction of admitted patients colonized	0.165
$\frac{1}{\delta_U}$	Length of stay, non-colonized patients	7
$\frac{1}{\delta_C}$	Length of stay, colonized patients	13
η	Hygienic level, HCW-patient	0.46
ξ	Hygienic level, volunteer-patient	0.23
β_{PH}	Patient-HCW contact rate	0.72
β_{PV}	Patient-volunteer contact rate	0.20
γ_H	HCW hand-washing rate	24
γ_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
ϕ	Fraction of admitted patients colonized	0.1
γ	Discharge rate, non-colonized patients	0.1
γ'	Discharge rate, colonized patients	0.05
μ	Staff decontamination rate	24
κ	Surfaces decontamination rate	1
β_{sp}	Staff-to-patient colonization rate	0.3
β_{se}	Staff-to-surface contamination rate	2
β_{ps}	Patient-to-staff contamination rate	2
β_{pe}	Patient-to-surface contamination rate	2
β_{es}	Surface-to-staff contamination rate	2
β_{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

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

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

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

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

	Meaning	Value
β_{AP1}	Patient-AP1 transmission rate	0.35
β_{AP2}	Patient-AP2 transmission rate	0.12
β_{Peri}	Patient-peripatetic transmission rate	0.07
μ	Hand-washing rate for all HCWs	1 - 24
γ^{-1}	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

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