

Magnus-type integrator for semilinear delay equations with an application to epidemic models

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Abstract

We consider a numerical method based on the Magnus series expansion, and show its second-order convergence when applied to a system of quasilinear delay equations. As an application, we take the delayed epidemic model and illustrate our results with numerical experiments.

Keywords Magnus method, convergence analysis, quasilinear delay equations, delayed epidemic models, SIR model

Epidemics affect everyone's lives. Due to frequent and long-distance traveling infectious diseases can spread rapidly, demanding more and more victims. Moreover, altered conditions caused by the climate change result in temporal and spatial changes in the source of infections. It is therefore indispensable that predictions of possible outbreaks are as accurate as possible. Besides collecting the data on past and present epidemics, the use of mathematical models offers a forecast capturing the main characteristics of an epidemics (such as the number of infected individuals). Since mathematical epidemic models are of a form of rather complex ordinary or partial differential equations, their exact solution cannot be determined. Instead, an approximation is computed by applying certain numerical methods.

The mathematical modeling of epidemics originates from the early twentieth century. A nice summary of the first attempts can be found in [14, Section 1.4]. Already Sir Ronald Ross, being awarded the Nobel Prize for the discovery of the malarial parasite, was convinced about the need of mathematical tools in epidemiology (see [14] and the references therein). The first epidemics models were proposed by Kermack and McKendrick in [9]. Their seminal work has lead to a large amount of completion and development of their model, making epidemic modeling a research field being promising in terms of social exploitation.

An important direction of developing epidemic models is the consideration of latent period, the time when a person is infected but is not infective, that is, the time from when the infected is really able to infect another individual. Incorporating the latent period leads to a system of differential equations with delay. In case of delayed epidemic models, the temporal change in the model's unknown quantities (usually the number of susceptible, infected, and recovered

individuals) do not only depend on their values at the actual time level but also on their values in the past (i.e. the latent period ago). Solution of delayed epidemic models needs efficient numerical methods which provides fast and accurate results.

In the present paper we propose a numerical method based on the use of Magnus method being originally developed for nonautonomous problems in [13]. We show that the delayed epidemic models can be written in a quasilinear form, and our approach leads to a positivity preserving and convergent numerical method which computes the numerical solution in an efficient way.

In Section 1 we introduce the Magnus-type integrator. Section 2 deals with the positivity preservation and convergence of Magnus-type integrator when applied to quasilinear delay equations. In Section 3 we use our results to treat delayed epidemic models. In Section 4 several numerical experiments are presented to illustrate our theoretical results.

1 Magnus-type integrator

In this section we introduce the Magnus-type integrator, and present two important convergence results from the literature which will be needed later on. For an arbitrary $d \in \mathbb{N}$, we consider the following nonautonomous evolution equation for the continuously differentiable unknown function $Y: [0, +\infty) \rightarrow \mathbb{R}^d$ where $A(t): \mathbb{R}^d \rightarrow \mathbb{R}^d$ is a linear operator for all $t \geq 0$ and $w \in \mathbb{R}^d$:

$$\begin{cases} Y'(t) = A(t)Y(t), & t \geq 0, \\ Y(0) = Y_0. \end{cases} \quad (1)$$

If problem (1) is well-posed, then for all $t \geq 0$ there exists a linear operator $\Omega(0 \rightarrow t): \mathbb{R}^d \rightarrow \mathbb{R}^d$, such that the exact solution has the form $Y(t) = e^{\Omega(0 \rightarrow t)}Y_0$ for all $t \geq 0$. We note that $(e^{\Omega(T_1 \rightarrow T_2)})_{T_1 \geq T_2}$ is an evolution family possessing the following properties (see e.g. in [17], [2]):

- (a) $e^{\Omega(T_2 \rightarrow T_3)}e^{\Omega(T_1 \rightarrow T_2)} = e^{\Omega(T_1 \rightarrow T_3)}$ for all $T_1 \leq T_2 \leq T_3 \in \mathbb{R}$,
- (b) the mapping $(T_1, T_2) \mapsto e^{\Omega(T_1 \rightarrow T_2)}$ is strongly continuous,
- (c) $\|e^{\Omega(T_1 \rightarrow T_2)}\| \leq Me^{\omega(T_2 - T_1)}$ holds for some $M \geq 1, \omega \in \mathbb{R}$ and all $T_1 \leq T_2 \in \mathbb{R}$.

An approximation of operator $\Omega(0 \rightarrow t)$ is based on the infinite series expansion of $Y(t)$ introduced in his seminal paper [13] by Magnus, and it reads for all $m \in \mathbb{N}$ as:

$$\begin{aligned} \Omega^{[0]}(0 \rightarrow t) &= 0, \\ \Omega^{[m]}(0 \rightarrow t) &= \int_0^t \sum_{k=0}^{\infty} \frac{B_k}{k!} \text{ad}_{\Omega^{[m-1]}(0 \rightarrow s)}^k A(s) ds, \end{aligned} \quad (2)$$

where $B_k, k \in \mathbb{N}_0$ are the Bernoulli numbers and $\text{ad}_{\Omega}^{[m]} A := [\Omega, \text{ad}_{\Omega}^{[m-1]} A]$ is the iterated commutator with $\text{ad}_{\Omega}^{[0]}$ being the identity. The corresponding approximate solution has the form

$$Y^{[m]}(t) = e^{\Omega^{[m]}(0 \rightarrow t)}Y_0 \quad (3)$$

and is called Magnus method. We cite the corresponding convergence result. From now on $C > 0$ denotes a generic constant.

Theorem 1 (Thm. 2.1 in [4]). *For the Magnus method (3), there exists a constant $C > 0$, being independent of t , such that $\|Y(t) - Y^{[m]}(t)\| \leq Ct^{m+1}$ holds for all $t \geq 0$.*

In what follows we consider the case $m = 1$, and use the notation $Y^{[1]}(t) =: y(t)$. Then the Magnus method (3) with $y_0 = Y_0$ is written as

$$y(t) = e^{\Omega^{[1]}(0 \rightarrow t)} y_0 \quad \text{with} \quad \Omega^{[1]}(0 \rightarrow t) = \int_0^t A(s) ds. \quad (4)$$

For further use, we give the general formula for all $T, \Delta T \geq 0$ real numbers:

$$\Omega^{[1]}(T \rightarrow T + \Delta T) = \int_0^{\Delta T} A(T + s) ds. \quad (5)$$

We note that the additivity of the integral implies the following relation:

$$\Omega^{[1]}(0 \rightarrow T + \Delta T) = \Omega^{[1]}(0 \rightarrow T) + \Omega^{[1]}(T \rightarrow T + \Delta T). \quad (6)$$

In order to define a numerical method later, we need another form of Magnus method (4). We define a time step $\tau > 0$ and the time levels $t_n = n\tau$ for all $n \in \mathbb{N}_0$. Then formulae (5) and (6) with $T = t_n$ and $\Delta T = \tau$ lead to the following form for the solution of Magnus method (4):

$$\begin{aligned} y(t_{n+1}) &= e^{\Omega^{[1]}(0 \rightarrow t_{n+1})} y_0 = e^{\Omega^{[1]}(t_n \rightarrow t_{n+1})} e^{\Omega^{[1]}(0 \rightarrow t_n)} y_0 = e^{\Omega^{[1]}(t_n \rightarrow t_{n+1})} y(t_n) \\ &= \exp \left(\int_0^\tau A(t_n + s) ds \right) y(t_n). \end{aligned} \quad (7)$$

Since the Magnus method (7) still consists of integrals, we need to approximate them by quadrature rules to get a numerical method. As before, we consider the time step $\tau > 0$ and the time levels $t_n = n\tau$, $n \in \mathbb{N}_0$. Then the first integral is approximated by the midpoint rule while the second by the left rectangle rule:

$$y(t_{n+1}) \approx e^{\tau A(t_n + \frac{\tau}{2})} y(t_n).$$

We denote the approximation of $y(t_n)$ at the time level $t_n = n\tau$ by \hat{y}_n for all $n \in \mathbb{N}_0$, and take $\hat{y}_0 = Y_0$. Then we obtain the **Magnus-type integrator** as follows:

$$\hat{y}_{n+1} = e^{\tau A(t_n + \frac{\tau}{2})} \hat{y}_n \quad (8)$$

with $y_0 = Y_0$. For further reference we cite here the results of González et al. reformulated for operators acting on the space \mathbb{R}^d .

Theorem 2 (Thm. 2. in [8]). *Suppose that the operator $A(t): \mathbb{R}^d \rightarrow \mathbb{R}^d$ is uniformly sectorial for $t \in [0, T]$. Thus, there exist constants $a \in \mathbb{R}$, $0 < \Gamma < \pi/2$, and $M_1 \geq 1$ such that $A(t)$ satisfies the resolvent condition*

$$\|(A(t) - \lambda I)^{-1}\| \leq \frac{M_1}{|\lambda - a|}$$

for any λ lying in the complement of the sector $S_\Gamma(a) = \{\lambda \in \mathbb{C}: |\arg(a - \lambda)| \leq \Gamma\} \cup \{a\}$. Suppose further that A is Lipschitz continuous, that is, there is a constant $M_2 > 0$ such that the estimate

$$\|A(t) - A(s)\| \leq M_2(t - s)$$

holds for all $0 \leq s \leq t \leq T$. Then there exists a constant $C > 0$, independent of n and τ , such that the error estimate

$$\|Y(t_n) - \hat{y}_n\| \leq C\tau^2(\|g'\|_\infty + \|g''\|_\infty)$$

holds with

$$\begin{aligned} g_n(t) &= (A(t) - A(t_n + \frac{\tau}{2}))Y(t) \\ \|g'\|_\infty &= \max_{n \in \mathbb{N}_0} \max_{t \in [t_n, t_{n+1}]} \|g'_n(t)\| \\ \|g''\|_\infty &= \max_{n \in \mathbb{N}_0} \max_{t \in [t_n, t_{n+1}]} \|g''_n(t)\| \end{aligned}$$

and $t_{n+1} \leq T$, whenever the right hand-side exists.

2 Magnus-type integrator for quasilinear delay equations

We present now how the Magnus-type integrator (13) can be applied to quasilinear delay equations, show that it preserves the positivity, and prove its second-order convergence.

Delay problems arise in numerous application fields where the system's temporal change depends on the system's past state(s) as well. They represent differential equations where the derivative does not only depend on the actual value of the unknown function but also on its values in the past. In what follows we treat quasilinear equations where operator $Q(w): \mathbb{R}^d \rightarrow \mathbb{R}^d$ is linear for all $w \in \mathbb{R}^d$, $\delta > 0$ is the time delay parameter, and $\phi: [-\delta, 0] \rightarrow \mathbb{R}^d$ is a given continuous function representing the history of the system. Then we are for the continuously differentiable unknown function $Y: \mathbb{R} \rightarrow \mathbb{R}^d$ satisfying the quasilinear delay equation of the form

$$\begin{cases} Y'(t) = Q(Y(t - \delta))Y(t), & t > 0, \\ Y(t) = \phi(t), & t \in [-\delta, 0]. \end{cases} \quad (9)$$

We note that for $\delta > 0$, the value $Y(t - \delta)$ is a given value for all $t \geq 0$. Therefore, the linear operator $Q(Y(t - \delta))$ is also known. The quasilinear delay equation (9) fits in the framework of problem (1) with the operator defined as

$$A(t) = Q(Y(t - \delta)) \quad (10)$$

for all $t \geq 0$. For the operator A defined in (10), an arbitrary $\tau > 0$, and $t_n = n\tau$, $n \in \mathbb{N}_0$, the Magnus method (7) has the form

$$y(t_{n+1}) = \exp\left(\int_0^\tau Q(Y(s - \delta)) ds\right)y(t_n),$$

where Y is given on the interval $[-\delta, 0]$. The exact solution Y can again be approximated by

the Magnus method (7) itself. Altogether we obtain

$$y(t_{n+1}) = \exp \left(\int_0^\tau Q(\tilde{y}_n(s)) ds \right) y(t_n) \quad \text{with}$$

$$\tilde{y}_n(s) = \begin{cases} \phi(t_n - \delta + s) & \text{for } t_n + s \in [0, \delta), \\ \exp \left(\int_0^s Q(\phi(t_n - 2\delta + \xi)) d\xi \right) y(t_n - \delta) & \text{for } t_n + s \in [\delta, 2\delta), \\ \exp \left(\int_0^s Q(y(t_n - 2\delta + \xi)) d\xi \right) y(t_n - \delta) & \text{for } t_n + s \geq 2\delta. \end{cases} \quad (11)$$

The Magnus-type integrator is derived by approximating the integral in (11) by midpoint rule:

$$y(t_{n+1}) \approx e^{\tau Q(\tilde{y}_n(\frac{\tau}{2}))} y(t_n).$$

And in the approximation of $\tilde{y}_n(\frac{\tau}{2})$ we use the left rectangle rule:

$$\tilde{y}_n(\frac{\tau}{2}) \approx \begin{cases} \phi(t_n - \delta + \frac{\tau}{2}) & \text{for } t_n \in [0, \delta), \\ e^{\frac{\tau}{2} Q(\phi(t_n - 2\delta))} y(t_n - \delta) & \text{for } t_n \in [\delta, 2\delta), \\ e^{\frac{\tau}{2} Q(y(t_n - 2\delta))} y(t_n - \delta) & \text{for } t_n \geq 2\delta. \end{cases}$$

In order to proceed, we choose an arbitrary number $N \in \mathbb{N}$, and define the time step as $\tau = \delta/N$. We suppose that τ satisfies the convergence criterion of Magnus method proved in [16, Thm. 3]:

$$\int_0^\tau \|A(t)\|_2 ds < \pi. \quad (12)$$

Then the **Magnus-type integrator** for quasilinear delay equation (9) has the form

$$y_{n+1} = e^{\tau Q(\tilde{y}_n)} y_n \quad \text{with}$$

$$\tilde{y}_n = \begin{cases} \phi(t_n - \delta + \frac{\tau}{2}) & \text{for } n = 0, 1, \dots, N-1, \\ e^{\frac{\tau}{2} Q(\phi(t_n - 2\delta))} y_{n-N} & \text{for } n = N, \dots, 2N-1, \\ e^{\frac{\tau}{2} Q(y_{n-2N})} y_{n-N} & \text{for } n \geq 2N. \end{cases} \quad (13)$$

In what follows we analyse the Magnus-type integrator (13) in terms of positivity preservation and convergence, moreover, in Section 3 we apply it to epidemic models.

In many physical/chemical/biological applications the unknown function should be positive (e.g. mass, pressure, concentration, population), unless one gets unreliable solutions. Thus, it is desirable that the numerical method preserves the sign of the solution, too. In what follows we give a sufficient condition for the positivity preservation of Magnus-type integrator (13).

Definition 3. (a) A vector having nonnegative elements only, is called a positive vector.

(b) A matrix $W \in \mathbb{R}^{d \times d}$ is called a Metzler matrix if its off-diagonal elements are nonnegative.

Remark 4. For $d \in \mathbb{N}$, let $W \in \mathbb{R}^{d \times d}$ be an arbitrary matrix and $w \in \mathbb{R}^d$ be a positive vector. Then [3, Lemma 5.3.a] states that $e^W w$ is a positive vector if and only if e^W has only nonnegative elements. Moreover, [3, Thm. 7.1] states that e^W has only nonnegative elements if and only if W is a Metzler matrix.

Corollary 5. For $d \in \mathbb{N}$, let $W \in \mathbb{R}^{d \times d}$ be an arbitrary matrix and $w \in \mathbb{R}^d$ be a positive vector. Remark 4 implies that $e^W w$ is a positive vector if and only if W is a Metzler matrix.

Proposition 6. Let $Q: \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$ be a function such that $Q(w)$ is a Metzler matrix for all positive vectors $w \in \mathbb{R}^d$. Then the Magnus-type integrator (13) preserves the positivity, that is, y_{n+1} is a positive vector for positive vectors y_n and $\phi(t)$, $t \in [-\delta, 0]$.

Proof. We remark first that if $Q(w)$ is a Metzler matrix then $tQ(w)$ is that as well for all $t \geq 0$. The Magnus-type integrator (13) has the form $y_{n+1} = e^{\tau Q(\tilde{y}_n)} y_n$. By Corollary 5 it suffices to show that the vector

$$\tilde{y}_n = \begin{cases} \phi(t_n - \delta + \frac{\tau}{2}) & \text{for } n = 0, 1, \dots, N-1, \\ e^{\frac{\tau}{2} Q(\phi(t_n - 2\delta))} y_{n-N} & \text{for } n = N, \dots, 2N-1, \\ e^{\frac{\tau}{2} Q(y_{n-2N})} y_{n-N} & \text{for } n \geq 2N \end{cases} \quad (14)$$

is positive for all $n \in \mathbb{N}_0$. We distinguish the following cases.

- (i) For $n = 0, \dots, N-1$, the vector $\tilde{y}_n = \phi(t_n - \delta + \frac{\tau}{2})$ is positive.
- (ii) For $n = N$, we have $\tilde{y}_N = e^{\frac{\tau}{2} Q(\phi(-\delta))} y_0$, where $\phi(-\delta)$ and y_0 are positive vectors by assumption. Thus, $\frac{\tau}{2} Q(\phi(-\delta))$ is a Metzler matrix and therefore \tilde{y}_N is a positive vector by Corollary 5.
- (iii) Let $n = N+1, \dots, 2N-1$. Then $\phi(t_n - 2\delta)$ is a positive vector and hence $\frac{\tau}{2} Q(\phi(t_n - 2\delta))$ is a Metzler matrix. Moreover, it holds that for indices

$$m := n - N - 1 = 0, \dots, N-2$$

the vector $y_{m+1} = e^{\tau Q(\tilde{y}_m)} y_{n-N} = e^{\tau Q(\phi(t_m - \delta + \frac{\tau}{2}))} \phi(t_m - \delta)$ is positive by step (i) and Corollary 5. Hence, y_{n-N} is a positive vector, too.

- (iv) For indices $n \geq 2N$ we proceed by induction. We saw that there was an index $k \in \mathbb{N}_0$ such that vector \tilde{y}_n was positive for all $n = 0, \dots, k$. Therefore the assumption on Q and Corollary 5 implies that

$$\text{vector } y_{n+1} \text{ is positive for all } n = 0, \dots, k. \quad (15)$$

Our aim is to show that \tilde{y}_{k+1} is a positive vector. The cases $k < 2N$ were shown in steps (i)–(iii), therefore, we consider the case $k \geq 2N$. Then formula (14) yields

$$\tilde{y}_{k+1} = e^{\frac{\tau}{2} Q(y_{k-2N})} y_{k-N}.$$

Since $n_1 := k - 2N - 1 < k$ and $n_2 := k - N - 1 < k$, the assertion (15) assures that the vectors $y_{n_1+1} = y_{n-2N}$ and $y_{n_2+1} = y_{n-N}$ are positive. Therefore, \tilde{y}_{k+1} is positive again by Corollary 5. Then the positivity of \tilde{y}_n for all $n \in \mathbb{N}_0$ follows by induction.

Since we obtained that \tilde{y}_n is a positive vector for all $n \in \mathbb{N}$, Corollary 5 implies the positivity of vector $y_{n+1} = e^{\tau Q(\tilde{y}_n)} y_n$ for all $n \in \mathbb{N}$, as well, which was to prove. \square

Our next aim is to analyse under which conditions the Magnus-type integrator (13) is convergent to the exact solution of problem (9). In what follows, for a function $F: \mathbb{R} \rightarrow \mathbb{R}^d$, the notation $F(\tau) = \mathcal{O}(\tau^p)$ means that there exists a constant $p > 0$ such that the relation

$$\lim_{\tau \rightarrow 0+} \frac{\|F(\tau)\|}{\tau^p} = 0$$

holds. We note that $F(\tau) = \mathcal{O}(\tau^p)$ implies that for all $\tau \in [0, T]$ there exists a constant $C > 0$ with $\|F(\tau)\| \leq C\tau^p$. We need a technical lemma.

Lemma 7. For matrices $W, Z \in \mathbb{R}^{d \times d}$ and $K := \max\{\|W\|, \|Z\|\}$, the following estimate holds:

$$\|e^W - e^Z\| \leq e^K \|W - Z\|.$$

Proof. By the matrix exponential and the telescopic identity we can write

$$\begin{aligned} \|e^W - e^Z\| &= \left\| \sum_{k=0}^{\infty} \frac{W^k}{k!} - \sum_{k=0}^{\infty} \frac{Z^k}{k!} \right\| = \left\| \sum_{k=0}^{\infty} \frac{1}{k!} (W^k - Z^k) \right\| \\ &= \left\| \sum_{k=0}^{\infty} \frac{1}{k!} \sum_{\ell=0}^{k-1} W^{k-1-\ell} (W - Z) Z^\ell \right\| \leq \sum_{k=0}^{\infty} \frac{1}{k!} \sum_{\ell=0}^{k-1} \|W\|^{k-1-\ell} \|W - Z\| \|Z\|^\ell \\ &\leq \sum_{k=0}^{\infty} \frac{1}{k!} \sum_{\ell=0}^{k-1} K^{k-1} \|W - Z\| = \sum_{k=0}^{\infty} \frac{k K^{k-1}}{k!} \|W - Z\| \\ &= \sum_{k=1}^{\infty} \frac{K^{k-1}}{(k-1)!} \|W - Z\| = \sum_{k=0}^{\infty} \frac{K^k}{k!} \|W - Z\| = e^K \|W - Z\| \end{aligned}$$

which was to prove. \square

Assumptions 8. Let $Q: \mathbb{R} \times \mathbb{R}^d \rightarrow \mathbb{R}^d$ be a function which satisfies the following.

- (a) The eigenvalues of matrix $Q(w) \in \mathbb{R}^{d \times d}$ lie in the complement of the sector $S_\Gamma(a) := \{\lambda \in \mathbb{C} : |\arg(\lambda - a)| \leq \Gamma\} \cup \{a\}$ with some $a \in \mathbb{C}$, $0 < \Gamma < \pi/2$, and for all $w \in \mathbb{R}^d$ positive vectors.
- (b) Function Q is Lipschitz continuous for positive vectors, that is, there exists a constant $L_Q \geq 0$ such that $\|Q(v) - Q(w)\| \leq L_Q \|v - w\|$ for all positive vectors $v, w \in \mathbb{R}^d$.
- (c) There exists a constant $M_Q \geq 0$ such that $\|Q(w)\| \leq M_Q$ for a bounded set of positive vectors $w \in \mathbb{R}^d$.
- (d) Function Q is twice differentiable with bounded derivatives for a bounded set of positive vectors.

Proposition 9. Under Assumptions 8, the Magnus-type integrator (13) is convergent of second-order, that is, there exists a constant $C > 0$, independent of τ and n , such that the error estimate $\|Y(t_n) - y_n\| \leq C\tau^2$ holds for all $n \in \mathbb{N}_0$ and $\tau \geq 0$ with $t_n = n\tau \in [0, T]$.

Proof. We first observe that the initial error vanishes. For all $n \in \mathbb{N}_0$, let \hat{y}_n be the solution defined in (8) with the operator A defined in (10):

$$\hat{y}_{n+1} = e^{\tau A(t_n + \frac{\tau}{2})} \hat{y}_n = e^{\tau Q(Y(t_n - \delta + \frac{\tau}{2}))} \hat{y}_n.$$

Since Assumptions 8 imply that the operator $A(t) = Q(Y(t - \delta))$ satisfies the assumptions of Theorem 2 with $a = 0$, we have

$$\|Y(t_{n+1}) - \hat{y}_{n+1}\| = \mathcal{O}(\tau^2). \quad (16)$$

The triangular inequality yields the following estimate on the global error:

$$\varepsilon_{n+1} := \|Y(t_{n+1}) - y_{n+1}\| \leq \|Y(t_{n+1}) - \hat{y}_{n+1}\| + \|\hat{y}_{n+1} - y_{n+1}\|, \quad (17)$$

where the first term on the right-hand side is $\mathcal{O}(\tau^2)$ by relation (16). Our aim is to estimate the second term on the right-hand side of (17). By using the telescopic identity we obtain the estimate

$$\begin{aligned}\|\widehat{y}_{n+1} - y_{n+1}\| &= \left\| \prod_{k=0}^n e^{\tau Q(Y(t_k - \delta + \frac{\tau}{2}))} Y_0 - \prod_{k=0}^n e^{\tau Q(\widetilde{y}_k)} Y_0 \right\| \\ &\leq \sum_{j=0}^n \prod_{k=0}^n \|e^{\tau Q(Y(t_k - \delta + \frac{\tau}{2}))}\| \|e^{\tau Q(Y(t_k - \delta + \frac{\tau}{2}))} - e^{\tau Q(\widetilde{y}_k)}\| \prod_{k=0}^n \|e^{\tau Q(\widetilde{y}_k)}\| \|Y_0\|,\end{aligned}$$

where \widetilde{y}_k was introduced in (13). By the assumption on the eigenvalues of $Q(w)$, we have $\|e^{tQ(w)}\| \leq 1$ for all $t \geq 0$ and positive vectors $w \in \mathbb{R}^d$. Thus, we can further write that

$$\|\widehat{y}_{n+1} - y_{n+1}\| \leq \sum_{j=0}^n \underbrace{\|e^{\tau Q(Y(t_k - \delta + \frac{\tau}{2}))} - e^{\tau Q(\widetilde{y}_k)}\|}_{(*)} \|Y_0\|, \quad (18)$$

where the term $(*)$ is a kind of local error to be estimated next. In what follows we use the exponential form of the matrices and the telescopic identity:

$$\begin{aligned} (*) &= \|e^{\tau Q(Y(t_k - \delta + \frac{\tau}{2}))} - e^{\tau Q(\widetilde{y}_k)}\| = \sum_{m=0}^{\infty} \frac{1}{m!} \|(\tau Q(Y(t_k - \delta + \frac{\tau}{2})))^m - (\tau Q(\widetilde{y}_k))^m\| \\ &\leq \sum_{m=0}^{\infty} \frac{1}{m!} \sum_{k=0}^{m-1} \|\tau Q(Y(t_k - \delta + \frac{\tau}{2}))\|^{m-1-k} \|\tau Q(Y(t_j - \delta + \frac{\tau}{2})) - \tau Q(\widetilde{y}_j)\| \|\tau Q(\widetilde{y}_k)\|^k \\ &\leq \sum_{m=0}^{\infty} \frac{1}{m!} \sum_{k=0}^{m-1} \tau^{m-1-k} M_Q^{m-1-k} \tau L_Q \underbrace{\|Y(t_j - \delta + \frac{\tau}{2}) - \widetilde{y}_j\|}_{\Delta_j} \tau^k M_Q^k \\ &= \sum_{m=0}^{\infty} \frac{\tau^m M_Q^{m-1}}{(m-1)!} L_Q \Delta_j = \tau e^{\tau M_Q} L_Q \Delta_j.\end{aligned}$$

Here we used Assumptions 8/(c) for the bounded sets $\{Y(t), t \in [0, T]\}$ and $\{\widetilde{y}_n, n \in \mathbb{N}_0 \text{ with } n\tau \in [0, T]\}$ of positive vectors. Since $e^{\tau M_Q} = \mathcal{O}(1)$, we have

$$(*) = \mathcal{O}(\tau) \Delta_j. \quad (19)$$

The term Δ_j can be computed from formulae (13) as

$$\Delta_j = \begin{cases} 0 & \text{for } j = 0, \dots, N-1 \\ \|Y(t_j - \delta + \frac{\tau}{2}) - e^{\frac{\tau}{2}Q(\phi(t_j - 2\delta))} y_{j-N}\| & \text{for } j = N, \dots, 2N-1 \\ \|Y(t_j - \delta + \frac{\tau}{2}) - e^{\frac{\tau}{2}Q(y_{j-N})}\| & \text{for } j \geq 2N. \end{cases}$$

Since the exact solution Y is given by an evolution family, we reformulate it as

$$\begin{aligned} Y(t_j - \delta + \frac{\tau}{2}) &= e^{\Omega(0 \rightarrow t_j - \delta + \frac{\tau}{2})} Y_0 = e^{\Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2})} e^{\Omega(0 \rightarrow t_j - \delta)} Y_0 \\ &= e^{\Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2})} Y(t_j - \delta). \end{aligned}$$

For $j = N, \dots, 2N-1$, we use the triangular inequality to get the estimate

$$\Delta_j \leq \underbrace{\|e^{\Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2})} - e^{\frac{\tau}{2}Q(\phi(t_j - 2\delta))}\|}_{(**)} \underbrace{\|Y(t_j - \delta)\|}_{\leq M_Y} + \underbrace{\|e^{\frac{\tau}{2}Q(\phi(t_j - 2\delta))}\|}_{\leq 1} \underbrace{\|Y(t_j - \delta) - y_{j-N}\|}_{\varepsilon_{j-N}}.$$

Here Y_M is the bound on the exact solution over the compact time interval $[0, T]$. It exists because Y is given by an evolution family which is strongly continuous, that is, Y is a continuous function over $[0, T]$, hence, bounded. To bound the term $(**)$, we use Lemma 7 and consider

$$\begin{aligned} & \left\| \Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2}) - \frac{\tau}{2} Q(\phi(t_j - 2\delta)) \right\| \\ & \leq \left\| \Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2}) - \int_0^{\tau/2} Q(\phi(t_j - 2\delta + s)) ds \right\| \\ & + \left\| \int_0^{\tau/2} Q(\phi(t_j - 2\delta + s)) ds - \frac{\tau}{2} Q(\phi(t_j - 2\delta)) \right\|. \end{aligned}$$

The definition (10) of operator A and formula (5) with the choice $T = t_n - \delta$, $\Delta T = \frac{\tau}{2}$ leads to

$$\int_0^{\tau/2} Q(\phi(t_j - 2\delta + s)) ds = \int_0^{\tau/2} A(t_j - \delta + s) ds = \Omega^{[1]}(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2}).$$

Hence, the first term on the right-hand side equals $\mathcal{O}(\tau^2)$ by Theorem 1. Since the second term corresponds to the local error of the left rectangle rule, it is of $\mathcal{O}(\tau^2)$, too. Altogether we have

$$\Delta_j \leq \mathcal{O}(\tau^2) + \varepsilon_{j-N} \quad \text{for all } j = N, \dots, 2N - 1. \quad (20)$$

For $j \geq 2N$, we proceed similarly:

$$\Delta_j \leq \underbrace{\left\| e^{\Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2})} - e^{\frac{\tau}{2} Q(y_{j-2N})} \right\|}_{(***)} \underbrace{\|Y(t_j - \delta)\|}_{\leq M_Y} + \underbrace{\left\| e^{\frac{\tau}{2} Q(y_{j-2N})} \right\|}_{\leq 1} \underbrace{\|Y(t_j - \delta) - y_{j-N}\|}_{\varepsilon_{j-N}}.$$

The term $(***)$ is approximated based on Lemma 7 as

$$\begin{aligned} & \left\| \Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2}) - \frac{\tau}{2} Q(y_{j-2N}) \right\| \\ & \leq \left\| \Omega(t_j - \delta \rightarrow t_j - \delta + \frac{\tau}{2}) - \int_0^{\tau/2} Q(Y(t_j - 2\delta + s)) ds \right\| \\ & + \left\| \int_0^{\tau/2} Q(Y(t_j - 2\delta + s)) ds - \frac{\tau}{2} Q(Y(t_j - 2\delta)) \right\| + \left\| \frac{\tau}{2} Q(Y(t_j - 2\delta)) - \frac{\tau}{2} Q(y_{j-2N}) \right\|. \end{aligned}$$

As before, the first term is of $\mathcal{O}(\tau^2)$ by Theorem 1, the second is the local error of the left rectangle rule being $\mathcal{O}(\tau^2)$ as well, and the third term is $\mathcal{O}(\tau)\varepsilon_{j-2N}$ by the Lipschitz continuity of Q . Altogether we have

$$\Delta_j \leq \mathcal{O}(\tau^2) + \varepsilon_{j-N} + \mathcal{O}(\tau)\varepsilon_{j-2N} \quad \text{for all } j \geq 2N. \quad (21)$$

Substitution of estimates (20) and (21) into formulae (19), (18), and (17) yields

$$\varepsilon_{n+1} \leq \mathcal{O}(\tau^2) + \mathcal{O}(\tau) \sum_{j=N}^n \varepsilon_{j-N} + \mathcal{O}(\tau^2) \sum_{j=2N}^n \varepsilon_{j-2N}$$

for all $n \in \mathbb{N}_0$. Hence, we have the following cases.

For $n = 0, \dots, N - 1$, we obtain $\varepsilon_{n+1} = \mathcal{O}(\tau^2)$.

For $n = N, \dots, 2N - 1$, we have

$$\varepsilon_{n+1} = \mathcal{O}(\tau^2) + \mathcal{O}(\tau) \sum_{j=N}^n \varepsilon_{j-N}$$

where we use the previous case to obtain

$$\varepsilon_{n+1} = \mathcal{O}(\tau^2) + \underbrace{\mathcal{O}(\tau)(n - N + 1)\tau}_{t_{n-N+1}} \mathcal{O}(\tau) = \mathcal{O}(\tau^2).$$

And finally for $n \geq 2N$, we use again the previous cases to obtain

$$\varepsilon_{n+1} = \mathcal{O}(\tau^2) + \mathcal{O}(\tau)t_{n-N+1}\mathcal{O}(\tau) + \mathcal{O}(\tau^2)t_{n-2N+1}\mathcal{O}(\tau) = \mathcal{O}(\tau^2).$$

Altogether we have $\varepsilon_{n+1} = \mathcal{O}(\tau^2)$ for all $n \in \mathbb{N}_0$ which was to show. \square

In what follows we apply the Magnus-type integrator (13) to delayed epidemic models.

3 Magnus-type integrator for delayed epidemic models

Let $S, I, R: \mathbb{R}_0^+ \rightarrow [0, 1]$ denote the number (or number ratio) of susceptible, infected, and recovered humans among the total population, respectively. Their temporal change depends on various phenomena, from which we only consider now the infection-related ones. Then the number of susceptible individuals decreases because they are in contact with infected people and get infected. More precisely, the actual change in $S(t)$ depends on $S(t)$ itself and on that with how many infected people they met the latent period ago, that is, at time $t - \delta$. The number of infected individuals naturally increases with the same amount, and decreases with the number of people who recover. Based on this consideration a compartment-type model can be formulated.

Let $\beta > 0$ denote the infection rate, $\gamma > 0$ the recovery rate, and $\delta > 0$ the latent period. Then for all $t > 0$, we consider the simplest but most used delayed epidemic model (based on Kermack–McKendrick [9] and Cooke [6], but see also in [5], [12], [15], [18], and the references therein):

$$\begin{cases} S'(t) = -\beta S(t) \frac{I(t - \delta)}{1 + \alpha I(t - \delta)}, \\ I'(t) = \beta S(t) \frac{I(t - \delta)}{1 + \alpha I(t - \delta)} - \gamma I(t), \\ R'(t) = \gamma I(t), \end{cases} \quad (22)$$

where the value of the parameter α depends on what kind of model we deal with:

$$\alpha = \begin{cases} 0 & \text{for bilinear incidence rate,} \\ 1 & \text{for saturated incidence rate.} \end{cases}$$

The incidence rate means the number of individuals who become infected in a unit of time. Initially, epidemic models were formulated by using the bilinear incidence rate (as in Kermack–McKendrick model [9]). The saturated incidence rate was introduced in [1] (see also in [7])

but without taking into account the latent period. It contains the crowding effect of infective individuals, and tends to a saturation value when the number of infected individuals gets large.

The initial condition to problem (22) reads as

$$S(0) = S_0, \quad I(0) = I_0, \quad R(0) = R_0, \quad \text{and} \quad I(s) = \varphi(s) \text{ for } s \in [-\delta, 0] \quad (23)$$

with $S_0, I_0, R_0 \geq 0$ given numbers and $\varphi: [-\delta, 0] \rightarrow \mathbb{R}_0^+$ given continuous function. We also assume that $\varphi(0) = I_0$.

Let $Y: \mathbb{R}_0^+ \rightarrow (\mathbb{R}_0^+)^3$ be defined as $Y(t) = (S(t), I(t), R(t))$ for all $t \geq 0$. Then the epidemic model (22) can be written as a quasilinear delay equation (9) with

$$Q(w) = \begin{pmatrix} -q(w^{(2)}) & 0 & 0 \\ q(w^{(2)}) & -\gamma & 0 \\ 0 & \gamma & 0 \end{pmatrix} \in \mathbb{R}^{3 \times 3} \quad \text{and} \quad q(w^{(2)}) = \frac{\beta w^{(2)}}{1 + \alpha w^{(2)}} \quad (24)$$

for any vector $w = (w^{(1)}, w^{(2)}, w^{(3)}) \in \mathbb{R}^3$. We recall that $I(t - \delta) = \varphi(t - \delta)$ is given for all $t \in [0, \delta]$.

Our aim is to apply the Magnus-type integrator (13) with Q defined in (24). Therefore, we split the interval $[-\delta, 0]$ into $N \in \mathbb{N}$ pieces, and define the time step as $\tau = \delta/N$ such that it satisfies condition (12). We note that the matrix function Q depends only on the second coordinate of the unknown function $y = (S, I, R)$. Therefore we can take $S_{n-N} = R_{n-N} = 0$ for $n < N$ without the loss of generality.

In what follows we prove the positivity preservation and the convergence of the Magnus-type integrator (13) with Q defined in (24).

Since epidemic models deal with population, it is natural to assume that the values of $S(t), I(t), R(t)$ are nonnegative for all time $t \geq 0$. Thus, we expect the same from the numerical method as well, that is, $S_n, I_n, R_n \geq 0$ should hold for all $n \in \mathbb{N}_0$.

Proposition 10. *The Magnus-type integrator (13) applied to delayed epidemic model (22) preserves the positivity.*

Proof. By Proposition 6 it suffices to show that $Q(w) \in \mathbb{R}^{3 \times 3}$ is a Metzler matrix for all positive vectors $w \in \mathbb{R}^3$. Since $\alpha, \beta, \gamma > 0$ holds and w is a positive vector, the off-diagonal elements of $Q(w)$ in (24) are nonnegative. \square

To prove the convergence we will need the following properties of the function Q .

Proposition 11. *Let $Q: \mathbb{R}^d \rightarrow \mathbb{R}^d$ be the matrix function defines in (24).*

- (a) *The eigenvalues of matrix $Q(w) \in \mathbb{R}^{3 \times 3}$ are nonpositive real numbers.*
- (b) *The matrix function Q is Lipschitz continuous for positive vectors.*
- (c) *The estimate $\|Q(w)\| \leq \beta w^{(2)} + \gamma$ holds for all positive vectors $w = (w^{(1)}, w^{(2)}, w^{(3)}) \in \mathbb{R}^3$.*
- (d) *For a bounded set $\{w_n \in (\mathbb{R}_0^+)^3, n \in \mathbb{N}\}$ we have $\|Q(w_n)\| \leq M_Q$ for all $n \in \mathbb{N}_0$ with some bound $M_Q \geq 0$.*
- (e) *The matrix function Q is twice differentiable and estimates $\|Q'(w)\| \leq \beta, \|Q''(w)\| \leq 2\alpha\beta$ hold for all positive vectors w .*

Proof. (a) Since matrix $Q(w)$, defined in (24), is a lower triangular matrix, its eigenvalues are the following:

$$\lambda_1 = 0, \quad \lambda_2 = -\gamma, \quad \lambda_3 = -\frac{\beta w^{(2)}}{1 + \alpha w^{(2)}}. \quad (25)$$

The positivity of vector w implies that $w^{(2)} \geq 0$, that is, all the eigenvalues above are nonpositive real numbers, hence, they lie in the complement of the desired sector.

- (b) Let $w = (w^{(1)}, w^{(2)}, w^{(3)}) \in \mathbb{R}^3$ and $z = (z^{(1)}, z^{(2)}, z^{(3)}) \in \mathbb{R}^3$ be arbitrary positive vectors. With functions Q and q defined in (24) we have

$$\begin{aligned} \|Q(w) - Q(z)\| &= \left\| \begin{pmatrix} -(q(w^{(2)}) - q(z^{(2)})) & 0 & 0 \\ q(w^{(2)}) - q(z^{(2)}) & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \right\| = |q(w^{(2)}) - q(z^{(2)})| \\ &= \left| \beta \left(\frac{w^{(2)}}{1 + \alpha w^{(2)}} - \frac{z^{(2)}}{1 + \alpha z^{(2)}} \right) \right| = \beta \left| \frac{w^{(2)} - z^{(2)}}{(1 + \alpha w^{(2)})(1 + \alpha z^{(2)})} \right|. \end{aligned}$$

Since $\alpha \in \{0, 1\}$ and $w^{(2)}, z^{(2)}$ are nonnegative real numbers, the denominator is greater than or equal to one. Thus we have the estimate

$$\|Q(w) - Q(z)\| \leq \beta \cdot |w^{(2)} - z^{(2)}| \leq \beta \cdot \max_{i=1,2,3} |w^{(i)} - z^{(i)}| = \beta \cdot \|w - z\|$$

proving the Lipschitz continuity of Q for positive vectors.

- (c) For a positive vector $w \in \mathbb{R}^3$ we have

$$\|Q(w)\| = \left\| \begin{pmatrix} -q(w^{(2)}) & 0 & 0 \\ q(w^{(2)}) & -\gamma & 0 \\ 0 & -\gamma & 0 \end{pmatrix} \right\| = q(w^{(2)}) + \gamma = \frac{\beta w^{(2)}}{1 + \alpha w^{(2)}} + \gamma \leq \beta w^{(2)} + \gamma$$

which was to be shown.

- (d) The previous statement (c) implies that for a bounded set $\{w_n \in (\mathbb{R}_0^+)^3, n \in \mathbb{N}\}$ we have $\|Q(w_n)\| \leq M_Q$ for all $n \in \mathbb{N}_0$ with some bound $M_Q \geq 0$.
- (e) Let $w = (w^{(1)}, w^{(2)}, w^{(3)}) \in \mathbb{R}^3$ be an arbitrary positive vector. Since Q maps from \mathbb{R}^3 into $\mathbb{R}^{3 \times 3}$, its derivative Q' can be represented as a continuous and linear mapping from \mathbb{R}^3 to \mathbb{R}^9 , which has only two non-zero elements $\pm q'(w^{(2)})$ and exists for all positive w vectors. The local boundedness follows from the nonnegativity of α and $w^{(2)}$:

$$\|Q'(w)\| = |q'(w^{(2)})| = \frac{\beta}{|(1 + \alpha w^{(2)})^2|} \leq \beta.$$

Similarly, for the second derivative we have the bound

$$\|Q''(w)\| = |q''(w^{(2)})| = \frac{2\alpha\beta}{|(1 + \alpha w^{(2)})^3|} \leq 2\alpha\beta.$$

These bounds were to prove. □

We show now the convergence of the Magnus-type integrator (13) when applied to the delayed epidemic model (22). We only need to check whether the assumptions in Proposition 9 are fulfilled for the function Q defined by formula (24).

Proposition 12. *The Magnus-type integrator (13) is convergent of second-order when applied to the delayed epidemic model (22).*

Proof. We have to check Assumptions 8.

- (a) Proposition 11/(a) implies that the eigenvalues of $Q(w)$ are nonpositive real numbers lying in the sector $S_\Gamma(0)$ with an arbitrary $0 < \Gamma < \pi/2$, for all positive vectors w .
- (b) The Lipschitz continuity of Q follows from Proposition 11/(b).
- (c) The boundedness of $Q(w)$ follows from Proposition 11/(d).
- (d) The differentiability and the boundedness of the derivatives follow from 11/(e).

The convergence follows now from Proposition 9. □

4 Numerical experiments

We illustrate our convergence result by numerical experiments done for the delayed epidemic model (22) with $\alpha = 1$. We remark that we use the approximation

$$\Phi(t_n - \delta + \frac{\tau}{2}) \approx \frac{1}{2}(\Phi(t_n - \delta) + \Phi(t_n - \delta + \tau)) \quad \text{for } n = 1, \dots, N-1$$

in formula (13).

In Figure 1 we present an example of the solution to the delayed epidemic model (22) with $\beta = 4$, $\gamma = 1$, $\tau = 0.01$. The upper panel shows the solution when the latent period is not taken into account (i.e. $\delta = 0$). In cases of the middle and lower panels the latent period is incorporated in the equations with the respective history functions

$$\varphi(t) = I_0, \tag{26}$$

$$\varphi(t) = I_0 + \frac{1}{2}s \tag{27}$$

in (23). One can see that already the constant history function changes the temporal behaviour of the solution: the maximal number of infected individuals is less than in the case without latent period. The linearly increasing history function introduces a new phenomena: the number of infected humans decreases at the beginning of the time interval.

The importance of considering the latent period is made clear by Figure 2 with $\beta = 1$, $\gamma = 1$ and the history function

$$\varphi(t) = I_0 - \frac{1}{2}s. \tag{28}$$

The upper panel shows the solution when the latent period is not taken into account (i.e. $\delta = 0$). In this case the model does not forecast any epidemic: the number of infected individuals monotonically decreases. However, when the latent period is incorporated in the equations, the

number of infected individuals starts to increase, that is, an epidemic occurs. The function I decreases only after the number S of susceptible humans reaches the critical value

$$S_c = \frac{\gamma}{\beta} \frac{1 + \alpha I(t_c - \delta)}{I(t_c - \delta)} I(t_c),$$

where t_c denote the time when $S(t_c) = S_c$. The difference between the two cases is caused by the delay. When there is a latent period $\delta > 0$, the actual derivative of I depends on how many susceptible humans there are now and how many infected humans there were the latent period ago because latter individuals begin to infect at present.

In Figure 3 the relative global error is shown for various values of time step τ . Since the exact solution to problem (22) is not known, we determine the relative global error $\varepsilon(\tau)$ by comparing to a reference solution y_{ref} computed by a small time step $\tau_{\text{ref}} = 10^{-4}$ as

$$\varepsilon(\tau) = \frac{\|y_{\text{ref}} - y_{n_{\text{max}}}\|}{\|y_{\text{ref}}\|}$$

where n_{max} refers to the index of the maximal time level. Since the results are presented in a logarithmic scale, the slope of the line fitted to them represents the numerical value of the convergence order. Since 1.97173 is approximately 2, the example illustrates well our second-order convergence result in Proposition 12.

Both the theoretical and the numerical results show that the Magnus-type integrator offers an efficient way of solving delayed epidemic models. Namely, it is a second-order method, so it gives a more accurate result than explicit or implicit Euler method which are usually used in population dynamics. Moreover, it is an explicit method, therefore, it does not need the solution of a system of nonlinear algebraic equations. We remark that the stability issue (which is usually solved by applying an implicit method) is not relevant here due to the exact computation of the matrix exponential. Therefore, the explicit kind of the method is not a disadvantage anymore, but on the contrary: it makes the computations fast and the implementation straightforward.

In a forthcoming paper we aim at generalizing our results concerning the second-order convergence of Magnus-type integrator to the case when it is applied to abstract semilinear delay equations corresponding to partial differential equations. We will work in the framework of linear evolution families which is an efficient tool for treating nonautonomous equations on appropriate function spaces.

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References

- [1] R. M. Anderson and R. M. May: Regulation and stability of host-parasite population interactions: I. Regulatory processes, *The Journal of Animal Ecology* **47** (1978) 219–267.
- [2] A. Bátkai, P. Csomós, B. Farkas: Operator splitting for nonautonomous delay equations, *Computers & Mathematics with Applications* **65** (2013) 315–324.
- [3] A. Bátkai, M. Kramar Fijavž, and A. Rhandi: *Positive Operator Semigroups, From Finite to Infinite Dimensions*, Birkhäuser, 2017.

- [4] F. Casas and A. Iserles: Explicit Magnus expansions for nonlinear equations, *Journal of Physics A: Mathematical and General* **39** (2006) 5445–5462.
- [5] V. Capasso and G. Serio: A generalization of the Kermack–Mckendrick deterministic epidemic model, *Math. Biosci.* **42** (1978) 41–61.
- [6] K. L. Cooke: Stability analysis for a vector disease model, *Rocky Mountain J. Math.* **9** (1979) 31–42.
- [7] L. Esteva and M. Matias: A model for vector transmitted diseases with saturation incidence, *Journal of Biological Systems* **9** (2001) 235–245.
- [8] C. González, A. Ostermann, M. Thalhammer: A second-order Magnus-type integrator for nonautonomous parabolic problems, *J. Comput. Appl. Math.* **189** (2006) 142–156.
- [9] W. Kermack and A. McKendrick: A contribution to mathematical theory of epidemics, *Proc. Roy. Soc. Lond. A* **115** (1927) 700–721.
- [10] W. Kermack and A. McKendrick: Contributions to the mathematical theory of epidemics–II. The problem of endemicity, *Bulletin of Mathematical Biology* **53** (1991)
- [11] W. Kermack and A. McKendrick: Contributions to the mathematical theory of epidemics–III. Further studies of the problem of endemicity, *Bulletin of Mathematical Biology* **53** (1991) 89–118.
- [12] W. Ma, M. Song, and Y. Takeuchi: Global stability of an SIR epidemic model with time delay, *Appl. Math. Lett.* **17** (2004) 1141–1145.
- [13] W. Magnus: On the exponential solution of a differential equation for a linear operator, *Comm. Pure Appl Math.* **7** (1954) 649–673.
- [14] M. Martcheva: *An Introduction to Mathematical Epidemiology*, Springer, 2015.
- [15] C. C. McCluskey: Global stability for an SIR epidemic model with delay and nonlinear incidence, *Nonlinear Analysis: Real World Applications* **11** (2010) 3106–3109.
- [16] P. C. Moan, J. Niesen: Convergence of the Magnus series, *J. Found. Comput. Math* **8** (2008) 291–301.
- [17] G. Nickel: Evolution semigroups for nonautonomous Cauchy problems, *Abstr. Appl. Anal.* **2** (1997) 73–95.
- [18] R. Xu and Z. Ma: Global stability of a SIR epidemic model with nonlinear incidence rate and time delay, *Nonlinear Analysis: Real World Applications* **10** (2009) 3175–3189.

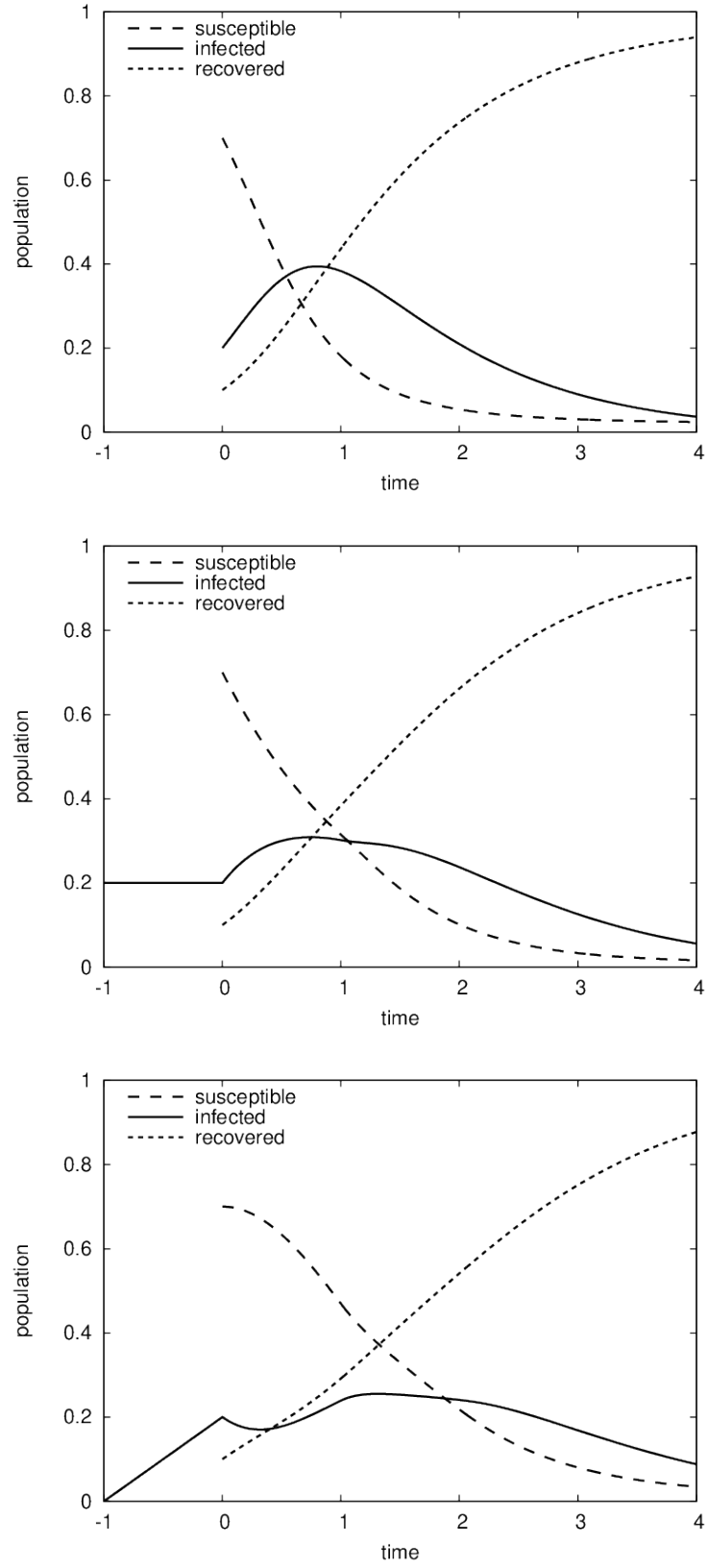


Figure 1: Numerical solution of delayed epidemic model (22) with $\alpha = 1$, $\beta = 4$, $\gamma = 1$, and $\tau = 0.01$ without taking into account the latent period (above) and with latent period and history functions (26) (middle) and (27) (below).

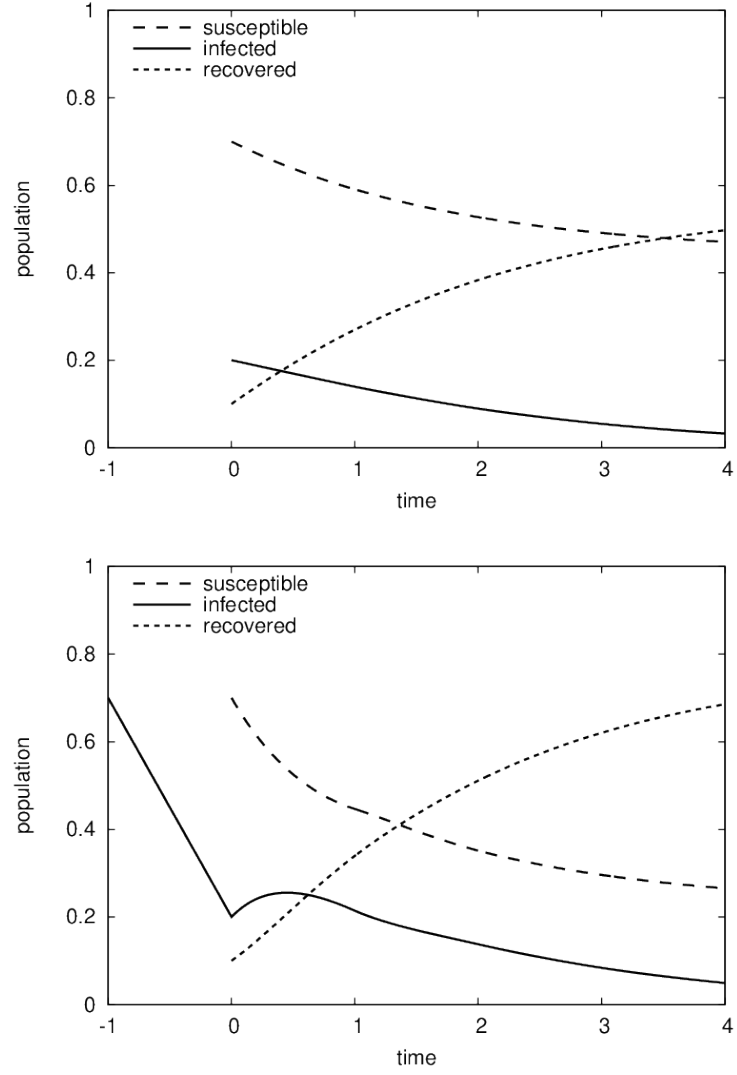


Figure 2: Numerical solution of delayed epidemic model (22) with $\alpha = 1$, $\beta = 1$, $\gamma = 1$, history function (28), and $\tau = 0.01$ without taking into account the latent period (above) and with latent period (below).

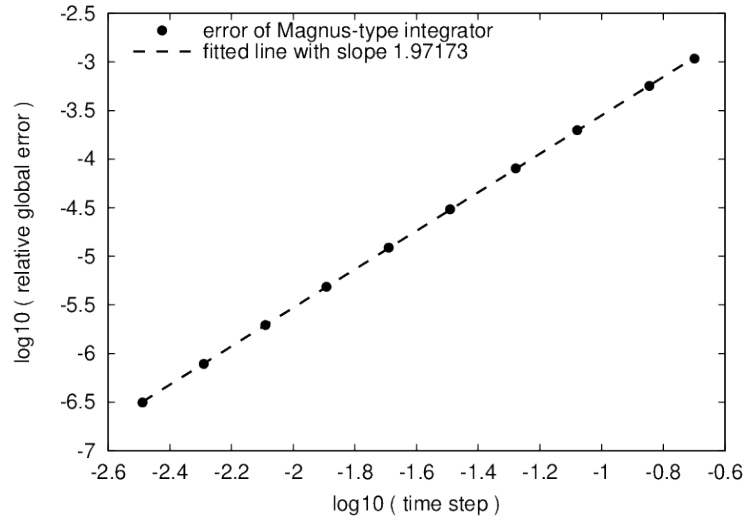


Figure 3: Order plot of Magnus-type integrator (13) when applied to the delayed epidemic model (22) with $\alpha = 1$, $\beta = 1$, $\gamma = 1$, $\tau_{\text{ref}} = 10^{-4}$ and history function (27). The dots corresponds to the relative global error of the model. The slope $1.97173 \approx 2$ of the fitted line represents the numerical value of the convergence order.