Dynamics of a Mathematical Hematopoietic Stem-Cell Population Model*

Daniel C. De Souza[†] and A. R. Humphries[‡]

- Abstract. We explore the bifurcations and dynamics of a scalar differential equation with a single constant delay which models the population of human hematopoietic stem cells in the bone marrow. One-parameter continuation reveals that with a delay of just a few days, stable periodic dynamics can be generated of all periods from about one week up to one decade! The long period orbits seem to be generated by several mechanisms, one of which is a canard explosion, for which we approximate the dynamics near the slow manifold. Two-parameter continuation reveals parameter regions with even more exotic dynamics including quasi-periodic and phase-locked tori, and chaotic solutions. The panoply of dynamics that we find in the model demonstrates that instability in the stem cell dynamics could be sufficient to generate the rich behavior seen in dynamic hematological diseases.
- Key words. delay-differential equations, hematopoietic stem cells, hematological diseases, canard explosion, torus, Lyapunov exponents

AMS subject classifications. 34K13, 34K18, 34K23, 34K60, 37G15, 37L30

DOI. 10.1137/18M1165086

1. Introduction. We study the dynamics and bifurcations of the delay-differential equation (DDE)

(1.1)
$$Q'(t) = -(\kappa + \beta(Q(t)))Q(t) + A\beta(Q(t-\tau))Q(t-\tau),$$

where $Q(t) \ge 0$ represents the concentration of hematopoietic stem cells (HSCs) in the bone marrow, $A \in (1, 2)$ is the amplification factor for cells undergoing division, $\tau > 0$ is the division time, the rate at which cells enter division, $\beta(Q)$, is a monotonically decreasing function of Q with $\lim_{Q\to\infty} \beta(Q) = 0$, and $\kappa > 0$ is the rate that the stem cells differentiate to the progenitors of circulating blood cells.

The DDE (1.1) represents the G_0 cell proliferation model of Burns and Tannock [11]. A full derivation of the DDE can be found in Mackey and Rudnicki [41], though it was first stated in the form (1.1) in Mackey [39]. In Bernard, Bélair, and Mackey [5], (1.1) was used to describe the stem cell dynamics as one component of a larger model describing the regulation

^{*}Received by the editors January 16, 2018; accepted for publication (in revised form) by J. Sieber January 30, 2019; published electronically April 16, 2019.

http://www.siam.org/journals/siads/18-2/M116508.html

Funding: The work of the first author was supported by National Council for Scientific and Technological Development of Brazil (CNPq) postdoctoral fellowship 201105/2014-4. The work of the second author was supported by a Discovery Grant from the Natural Science and Engineering Research Council (NSERC), Canada.

[†]Department of Mathematics and Statistics, McGill University, Montreal, Quebec, H3A 0B9, Canada. Current address: Institute of Immunology and Infection Research, University of Edinburgh, Ashworth Labs, Edinburgh, EH9 3JT, Scotland (daniel.desouza@ed.ac.uk).

[‡]Department of Mathematics and Statistics, McGill University, Montreal, Quebec, H3A 0B9, Canada, and Department of Physiology, McGill University, Montreal, Quebec, H3G 1Y6, Canada (tony.humphries@mcgill.ca, http://www.math.mcgill.ca/humphries).

of circulating neutrophil concentrations. Since then many mathematical models have appeared which contain (1.1), or a variant, within larger hematopoiesis models for the production and regulation of neutrophils, erythrocytes, and platelets [2, 15, 16, 18, 37]. Multiple versions of (1.1) have also been coupled together to model discrete levels of stem cell maturity [1, 51].

The human hematopoietic system produces about 10¹¹ blood cells of various types per day [32], of which erythrocytes (red blood cells), neutrophils (a type of white blood cell), and platelets are the most common, in a production process which is tightly regulated by myriad feedback loops. In dynamical diseases including cyclic neutropenia (CN), cyclic thrombocytopenia (CT), and periodic chronic myelogenous leukemia (PCML), oscillations are observed in the circulating concentrations of one or more of the cell lines [23].

In mathematical models of hematopoiesis these oscillations arise through Hopf bifurcations as one or more parameters are varied in the model. Two principle mechanisms have been proposed to drive the oscillations in different dynamical diseases [5]. There can be an instability in the production of the HSCs themselves, with the oscillating HSC numbers then leading to oscillations in the concentrations of peripheral cells. This occurs in PCML where leukemic HSCs typically present a chromosome abnormality [49]. An alternative mechanism is that oscillations in one cell line can be created through an abnormality in the production of precursor cells in that cell lineage, with the feedback loops from that lineage causing oscillations in the numbers of HSCs differentiating into other cell lines creating concomitant oscillations in the other cell lineages. This occurs in CN, for which a mutation in the ELANE gene that encodes neutrophil elastase leads to increased apoptosis in the neutrophil progenitor cells during mitosis [19].

In other dynamical diseases it remains an open question whether the oscillations are driven by an inherent instability in the HSCs or whether an instability in production of one of the blood cell lineages is creating the oscillations seen in circulating concentrations. The second possibility is difficult to investigate directly, due to the complexity of the hematopoietic models, with, for example, the granulopoiesis model of Craig, Humphries, and Mackey [18] having five equations, over 20 parameters, and state-dependent delays. In the current work, motivated by the first possibility, we investigate the dynamics of the simple HSC model (1.1) as parameters are varied and explore the dynamics that arise. Such an approach alone will not definitively answer the question of whether the oscillations in specific dynamical diseases are driven by inherent instability in the HSC dynamics. However, (1.1) is often incorporated in more complicated hematopoietic models, and if the HSCs can oscillate in the decoupled equation (1.1), these oscillations could drive oscillations in the production rates of the mature circulating blood cells which are all produced from the HSCs. Thus, studying the dynamics of (1.1) allows us to determine when instabilities in the HSC dynamics may arise, and hence if it is feasible for these to drive oscillations in the circulating blood cells concentrations.

Although (1.1) has been used and studied in numerous models, the codimension-one bifurcation analysis is incomplete, and little is known about codimension-two bifurcations. In section 2 we review the model (1.1) and its basic dynamical properties including existence and positivity of solutions, nondimensionalized formulation, homeostasis (the stable state of an organism maintained by physiological processes) parameter values, and existence and stability of steady states. In section 2.1 we discuss the stability boundary of the steady state with respect to the delay τ . In section 3 we carry out a numerical bifurcation analysis of (1.1) by performing parameter continuation on solutions as three of the parameters that control the dynamics are varied individually or pairwise. The one-parameter continuations reveal sub- and supercritical Hopf bifurcations, fold bifurcations of periodic orbits, and period-doubling bifurcations. This results in bistability between a stable periodic orbit and a steady state and bistability of two periodic orbits. We also find limit cycles of periods ranging from a week to over 9 years, and an apparent canard explosion [4]. The two-parameter continuations allow us to map out the curves of Hopf, period-doubling, and fold bifurcations to determine regions of parameter space for which interesting dynamics occur, and also reveal torus (or Neimark–Sacker) bifurcations.

In the following sections we explore some of the more interesting dynamics in more detail. In section 4 we study a canard explosion for which the period of solutions increases from about 50 days to over 700 days over an exponentially small parameter interval. We show how to approximate the slow manifold associated with these solutions and show that this manifold has both stable and unstable components. In section 5 we consider nonperiodic and chaotic solutions. First in section 5.1 we investigate the torus bifurcations found in section 3 and find a stable invariant torus in the dynamics. We compute Lyapunov exponents and a Poincaré section to show that the dynamics do indeed correspond to a quasi-periodic orbit which envelops the unstable periodic orbit from which the torus bifurcated. We also find parameter values for which there is phase locking on the torus and present the resulting stable periodic orbits. In section 5.2 we study the dynamics between the period doublings and find period-doubling cascades leading to chaos. The chaotic nature of the dynamics is verified numerically by showing that the leading Lyapunov exponent is positive and by visualizations of the attractor which reveal some of its fractal structure. Parameter continuation in opposite directions reveals hysteresis with parameter intervals for which stable chaotic dynamics can coexist with a stable periodic orbit, or even coexist with a second chaotic attractor. We also find parameter values for which there appears to be transient chaos. In section 5.3 we present an example of a branch of periodic orbits which snakes in parameter space resulting in a small parameter region in which over 50 limit cycles coexist. Period-doubling cascades either side of this region lead to additional parameter regions of chaotic dynamics.

In section 6 we discuss the physiological plausibility and implications of the results with regard to periodic hematological disorders, and section 7 includes further discussion and conclusions.

2. Hematopoietic stem cell equation. HSC dynamics can be described by the classic G_0 cell cycle model of Burns and Tannock [11]. The HSCs are distinguished between two phases, the proliferating phase and the resting or G_0 phase. We denote the concentration of HSCs in the resting phase by Q. From the resting phase HSCs may enter the proliferating phase at a rate $\beta(Q)$, or differentiate at a constant rate κ , or remain in the resting phase. Once HSCs enter the proliferating phase they are lost by apoptosis with a constant rate γ or undergo mitosis. The time to complete the cell cycle is τ . After mitosis cells return to the G_0 resting phase, from whence the cycle may begin again. In the resting phase HSCs are quiescent, while in the proliferating phase they are active and distinguished between four subphases: G_1 , S, G_2 and M. Cells at gap G_1 increase in size and are committed to go through the cell cycle

DYNAMICS OF A HEMATOPOIETIC STEM-CELL MODEL



Figure 1. Schematic representation of the classical G_0 model for HSCs. The proliferating phase of the cell cycle is divided between four subphases: gap one G_1 , synthesis S, gap two G_2 , and mitosis M. Cells in the resting phase (gap zero G_0) may differentiate with rate κ or enter the cell cycle with rate $\beta(Q)$. Cells in the proliferation phase may be lost by apoptosis with rate γ ; otherwise they reenter the resting phase after mitosis, τ time units after they left the resting phase.

and undergo mitosis. At S phase DNA synthesis occurs, at gap G_2 cells continue to grow, while in the mitotic phase M cells stop growing and undergo cell division.

A schematic of the model is presented in Figure 1. Following Mackey [39, 41] this model can be stated as the DDE (1.1) where $\beta(Q)$ is a Hill function defined by

(2.1)
$$\beta(Q) = f \frac{\theta^s}{\theta^s + Q^s},$$

and the HSC amplification factor A is given by

The parameters κ , γ , τ , θ , f, and s are all strictly positive, and we are interested in nonnegative solutions $Q(t) \ge 0$, since Q(t) represents a blood cell population.

Aspects of the dynamics of the HSC model (1.1) have been studied by a number of authors, mainly concentrating on the existence and stability of the steady states, and one-parameter continuation of some of the periodic orbits that arise [3, 6, 41]. A special case of (1.1) with $\beta(Q)$ replaced by a step function, corresponding to the limit as $s \to \infty$ in (2.1), allows explicit stable periodic solutions to be constructed [49, 50]. The existence of stable periodic solutions for s large was subsequently established [40], as a perturbation of the $s = \infty$ solutions. In contrast, we will study the dynamics of (1.1) with $\beta(Q)$ a Hill function, with s small, as is usually considered to be the case in hematopoiesis models.

In order to solve the DDE (1.1) for $t \ge 0$ it is necessary to define an initial function $Q(t) = \varphi(t)$ for $t \in [-\tau, 0]$. For $\varphi \in C \coloneqq C([-\tau, 0], [0, \infty))$, that is continuous, bounded, and nonnegative, from the following theorem the solution of (1.1) is also bounded and nonnegative. It follows that the DDE (1.1) can be considered as an infinite-dimensional dynamical system with phase space C [54].

Theorem 2.1. If $Q(t) = \varphi(t)$ for $t \in [-\tau, 0]$ where $\varphi \in C([-\tau, 0], [0, \infty))$, then (1.1) has a unique solution Q(t) defined for all $t \ge 0$ and which satisfies $Q(t) \in [0, M]$ for all $t \ge 0$ for some $M < \infty$.

Proof. As already noted by other authors, uniqueness and local existence of solutions follow from the method of steps. It suffices to show that solutions are bounded to obtain global existence and complete the proof [40, 49, 50].

To show positivity of the solution, let $t^* \ge 0$ such that $Q(t) \ge 0$ for $t \in [-\tau, t^*]$. It follows easily from (1.1) that Q'(t) > 0 for all $t \in (t^*, t^* + \tau)$ for which $Q(t) \in (-\theta, 0)$. This leads to a contradiction unless $Q(t) \ge 0$ for all $t \in (t^*, t^* + \tau)$. Hence $Q(t) \ge 0$ for all $t \ge 0$. The existence of an upper bound $Q(t) \le M$ follows from $\limsup_{t\to\infty} Q(t)$ being bounded, which was shown by Mackey and Rudnicki [41] for nonnegative solutions with a general class of monotonic functions $\beta(Q)$ that includes (2.1).

In (1.1)–(2.2) there are six parameters $\{\kappa, \gamma, \tau, \theta, f, s\}$, but we can reduce these to four by nondimensionalizing the equations. Let $\hat{t} \coloneqq t/\tau$, $\hat{f} \coloneqq \tau f$, $\hat{\kappa} \coloneqq \kappa/f$, $\hat{\gamma} \coloneqq \gamma \tau$, and $\hat{Q}(\hat{t}) \coloneqq Q(t)/\theta$; then $\hat{Q}(\hat{t}-1) = Q(t-\tau)/\theta$. It is also convenient to define $\hat{A} \coloneqq 2e^{-\hat{\gamma}} = 2e^{-\gamma\tau} = A$ and notice that $\beta(\hat{Q}) = f/(1+\hat{Q}^s)$. Then (1.1) becomes

(2.3)
$$\frac{1}{\hat{f}}\frac{d\hat{Q}}{d\hat{t}}(\hat{t}) = -\hat{\kappa}\hat{Q}(\hat{t}) - \frac{\hat{Q}(\hat{t})}{1+\hat{Q}(\hat{t})^s} + \hat{A}\frac{\hat{Q}(\hat{t}-1)}{1+\hat{Q}(\hat{t}-1)^s},$$

which depends on the four parameters \hat{f} , $\hat{\kappa}$, s, and \hat{A} (or $\hat{\gamma}$). Although many mathematicians would prefer to study the nondimensionalized DDE (2.3) instead of (1.1), we chose to work with (1.1) so that the solutions and bifurcations that we find have direct physiological interpretations. But, as suggested by the nondimensionalization, we need only vary four parameters in (1.1). It is easily seen that varying the four parameters γ , κ , τ , and s in (1.1), with the other parameters held constant, we can reproduce all possible values of \hat{f} , $\hat{\kappa}$, s, and \hat{A} in (2.3), and hence we will only need to consider the variation of parameters from among these four.

In Table 1 we state homeostatic values of the parameters for the model (1.1)-(2.2). The values in Table 1 are all taken from Craig, Humphries, and Mackey [18]. The first two parameters in the table, Q^h the homeostatic concentration of HSCs, and $\beta(Q^h)$ the homeostatic rate that cells enter the cell cycle, do not appear explicitly in the model (1.1)-(2.2) but are used to calculate the last two parameters. To ensure that Q^h is exactly the homeostatic concentration

Table 1

Homeostasis values of the parameters for the mathematical granulopoiesis model (1.1)-(2.2), which are all taken from Craig, Humphries, and Mackey [18]. The last three parameters with values denoted by * are only stated to four significant figures here, but actual values are computed to full double precision in MATLAB [45] using (2.2) and (2.4).

Name	Interpretation	Value	Units
Q^h	HSC homeostasis concentration	homeostasis concentration 1.1	
$\beta(Q^h)$	Homeostasis cell cycle entry rate	0.043	$days^{-1}$
γ	HSC apoptosis rate	0.1	$days^{-1}$
au	Time for HSC reentry	2.8	days
f	Maximal HSC reentry rate	8	$days^{-1}$
s	HSC reentry Hill coefficient	2	—
A	HSC amplification factor	1.512^{*}	—
θ	Half-effect HSC concentration	0.08086^{*}	10^6 cells/kg
κ	HSC differentiation rate to all lines	0.022^{*}	$days^{-1}$

of HSCs in the model (1.1)-(2.2), the last three parameters are computed in double precision. While A is given directly by (2.2), rearranging (1.1) and (2.1) at homeostasis implies that

(2.4)
$$\theta = Q^h \left(\frac{f}{\beta(Q^h)} - 1\right)^{-1/s}, \qquad \kappa = (A-1)\beta(Q^h)$$

We will use the parameters of Table 1 as a starting point for our bifurcation studies.

For general parameter values, from (1.1), steady states satisfy

(2.5)
$$0 = [(A-1)\beta(Q) - \kappa]Q.$$

Hence the DDE (1.1) has the trivial steady state Q(t) = 0 for all values of the parameters. The trivial steady state has been shown to be globally asymptotically stable if $\kappa > f(A-1)$ [41]. Equation (2.5) has another solution Q^* which satisfies

(2.6)
$$\beta(Q^*) = \frac{\kappa}{(A-1)}.$$

Since $\beta(Q)$ is monotonic, this defines a unique nontrivial steady state Q^* . Using (2.1) we have

(2.7)
$$Q^* = \theta \left[f \frac{(A-1)}{\kappa} - 1 \right]^{1/s}$$

From (2.7), along with the relation (2.2) we obtain that $Q^* > 0$ if and only if the upper bounds

(2.8)
$$\kappa < f(A-1), \qquad \gamma \tau < \ln\left(\frac{2f}{\kappa+f}\right),$$

on the parameters κ , γ , and τ are satisfied. In the rest of this work, we will consider the case where (2.8) holds and there are two steady states, Q = 0 and $Q^* > 0$. From (2.2) and (2.8) we require $A \in (1 + \kappa/f, 2)$ for $Q^* > 0$ to exist. Regarding the steady state $Q^* > 0$ as a function of the differentiation rate κ , the death rate γ , or the cell cycle duration τ , it is easy to see from (2.7) that Q^* is a monotonically decreasing function with respect to each of these parameters, as would be expected from a physiological point of view. Furthermore $Q^* \to 0$ as equality is approached in (2.8). We denote the nonzero steady state by Q^h only when the parameters take their homeostasis values from Table 1, and by Q^* otherwise. Unless otherwise stated, all the values in this paper are given in the same units as in Table 1.

2.1. Stability boundary. To determine the stability of the steady state Q^* , linearize the DDE (1.1) around Q^* with $z(t) \coloneqq Q(t) - Q^*$ to get

(2.9)
$$z'(t) = az(t) + bz(t - \tau).$$

It is convenient to define

$$h(x) \coloneqq x\beta(x);$$

then the parameters a and b can be written as

(2.11)

$$a = -\kappa - \beta(Q^*) - Q^* \beta'(Q^*) = -\kappa - h'(Q^*), \qquad b = A\left(\beta(Q^*) + Q^* \beta'(Q^*)\right) = Ah'(Q^*)$$

Seeking a nontrivial solution $z(t) = z_0 e^{\lambda t}$ for (2.9), we get the characteristic equation

(2.12)
$$p(\lambda) \coloneqq \lambda - a - be^{-\lambda\tau} = 0,$$

first studied by Hayes [28]. According to the principle of linearized stability [54] the stability analysis of the steady states for the nonlinear DDE (1.1) is reduced to the stability analysis of the steady state of the linearized equation (2.9). Stability analysis of (2.9) is a standard example in DDEs and can be found in [8, 26, 29, 54]. The steady state is unstable if b > -awith a characteristic value $\lambda > 0$, and it is asymptotically stable if a < -|b| < 0, which is sometimes called the delay-independent stability region [33]. The interesting parameter region is for $b \leq -|a| < 0$, where the steady state is asymptotically stable for

and unstable if $\tau > \tau_1(a, b)$. The curve C_0 described by $\tau_1(a, b)$ is contained in the region $a \leq 1/\tau$ and $b \leq -1/\tau$ and can be parameterized [54] as

(2.14)
$$C_0 = \left\{ (a,b) = (\omega \cot(\omega)/\tau, -\omega \csc(\omega)/\tau), \quad \omega \in [0,\pi) \right\}.$$

On this curve the characteristic equation (2.12) has an imaginary solution $\lambda = \pm i\omega$. The parameter region $(a\tau, b\tau) \in \mathbb{R}^2$ for which Q^* is stable is illustrated in Figure 2.

In the context of the DDE (1.1) the trivial steady state Q = 0 is unstable when $Q^* > 0$ [51]. For the stability of Q^* , from (2.6) and (2.11) we have $a + b = (A - 1)Q^*\beta'(Q^*)$, and hence using (2.1) and (2.7) we find that $-s\kappa < -s\kappa(1 - \kappa/[f(A - 1)]) = a + b < 0$. Thus when $Q^* > 0$ we have a + b < 0, and Q^* can only lose stability if (a, b) crosses the curve C_0 as parameters are varied.

The steady state $Q^* > 0$ is stable for all τ sufficiently small, since $\lambda = -(a+b) < 0$ when $\tau = 0$. It is also stable for all τ sufficiently large. This follows from noting that b > 0 when $h'(Q^*) > 0$, which holds whenever $\tau > \tau_2$ where τ_2 is defined by (2.16) below. Then, for $\tau > \tau_2$ the parameters (a, b) are in the upper half of the delay-independent stability region a < -|b| < 0.



Figure 2. The parameter regions for which the steady state of (2.9) is stable (white) and unstable (shaded). Also shown (red) is the locus in (a, b) of the parameters defined by (2.11) as τ is varied with the other parameters all at their values from Table 1. The values of τ from (2.15) are indicated on this curve.

DYNAMICS OF A HEMATOPOIETIC STEM-CELL MODEL

The steady state Q^* may lose stability in a Hopf bifurcation if the parameters cross the curve C_0 in the (a, b) parameter space, but since the steady state is stable for τ small and for τ large, such Hopf bifurcations will occur in pairs, corresponding to crossings of C_0 in opposite directions. For many DDEs the steady state gains ever more characteristic values with positive real part as the delay τ is increased [46]. The Burns–Tannock DDE (1.1) does not behave like that because of the exponential term in $A = 2e^{-\gamma\tau}$ representing mortality during the cell cycle.

For the homeostasis parameter values in Table 1 we have $f(A-1) = 4.093 > \kappa = 0.022$ so (2.8) is satisfied and there exists a unique positive steady state, $Q^h = 1.1 > 0$. Furthermore, at homeostasis a = 0.020540 and b = -0.064298 and $\tau = 2.8$ which is inside the stability region, as illustrated in Figure 2, so the homeostasis steady state Q^h is asymptotically stable.

If τ is varied and all the other parameters are at their homeostasis values from Table 1, we find that $Q^* > 0$ is stable for $\tau \in [0, \tau_1^-)$ and for $\tau \in (\tau_1^+, \tau_{\max})$ where

$$(2.15) \quad 0 < \tau^{h} = 2.8 < \tau_{1}^{-} = 5.74851 < \tau_{1}^{+} = 6.87437 < \tau_{2} = 6.87662 < \tau_{\max} = 6.90401.$$

Here, τ_{max} is given by (2.8), while

(2.16)
$$\tau_1^{\pm} = \frac{\cos^{-1}((\kappa + h'(Q^*))/Ah'(Q^*))}{\sqrt{(Ah'(Q^*))^2 - (\kappa + h'(Q^*))^2}}, \qquad \tau_2 = \frac{1}{\gamma} \ln\left(\frac{1}{2} \left[1 + \frac{\kappa s}{f(s-1)}\right]\right)^{-1}$$

The formula for τ_2 is found by using the expressions for h(Q) and Q^* and solving for b = 0. This formula was already stated by Pujo-Menjouet and others [49, 50], where it was erroneously claimed that $\tau = \tau_2$ was a stability boundary. As noted above, and as shown in Figure 2, the parameters corresponding to $\tau = \tau_2$ are in the interior of the stability region.

The expression for τ_1^{\pm} in (2.16) is obtained from (2.13) on substituting the values from (2.11). The two values for τ_1 correspond to the parameters (a, b) defined by (2.11) crossing the curve C_0 twice as τ is increased. The equation $\tau = \tau_1$ from (2.13) can have two solutions, because in the expression for τ_1 the value of A itself depends on τ , making the equation $\tau = \tau_1$ implicit in τ . The locus of the parameters (a, b) and the corresponding τ values from (2.15) are illustrated in Figure 2.

If parameters are varied so that τ crosses the boundary τ_1 , either by varying τ itself or by varying parameters to change the values of τ_1 , then the stability of Q^* changes and a periodic orbit with period $2\pi/\omega$ is created in a Hopf bifurcation. The periodic orbits thus created have been explored to some extent [24, 40, 49, 50], but the details of the Hopf bifurcation and its normal form have only been studied more recently [51].

Because (2.12) has infinitely many roots, it is possible for additional pairs of complex conjugate characteristic values to cross the imaginary axis resulting in additional Hopf bifurcations. These will occur on curves C_{2n} , which are defined by (2.14) but for $\omega \in [2n\pi, (2n+1)\pi)$ [54]. None of these curves intersect, so there are no double-Hopf bifurcations in which two pairs of characteristic values cross the imaginary axis at the same time.

3. Bifurcations of the HSC equation. In this section we vary parameters from their homeostasis values so that the perturbed steady state Q^* becomes unstable, and we survey the bifurcations and dynamics that arise. We begin our numerical bifurcation analysis of the

DDE (1.1), by studying codimension-one bifurcations performing parameter continuation on τ , γ , and κ , one at a time. We will also study codimension-two bifurcations carrying out a two-parameter continuation for each pair of these three parameters. Our numerical bifurcation diagrams are constructed using the well-established DDEBiftool package [21, 53], which runs under MATLAB [45]. This software finds periodic orbits using a boundary value approach and is able to find stable and unstable solutions and continue the solutions as parameters are varied and detect stability and bifurcations.

As noted after (2.3), the dynamics of the HSC DDE (1.1) only depend on the four parameters τ , γ , κ , and s. The dependence of the dynamics on the parameter s has previously been studied for small integer values of s (through numerical simulation) and analytically in the limit as $s \to \infty$ (in which case the Hill function (2.1) simplifies to a Heaviside function) [40, 49, 50]. In the current work we will keep s = 2 fixed and equal to its value in Craig, Humphries, and Mackey [18], and we will only vary the parameters τ , γ , and κ . In the following, unless mentioned otherwise, all parameters take the values stated in Table 1. Recall that the homeostasis steady state Q^h is stable.

3.1. One-parameter continuation. We begin by varying the differentiation rate κ (with all the other parameters held constant) with the bifurcation diagram presented in Figure 3. The nontrivial steady state Q^* (given by (2.7)) is seen to be unstable for an interval of κ values between two Hopf bifurcation points and stable for κ outside this interval. The Hopf bifurcation point at $\kappa \approx 0.17632$ is subcritical leading to a branch of unstable periodic orbits which becomes stable at $\kappa \approx 0.16872$ in a fold bifurcation of periodic orbits. This creates a small interval of bistability between the stable steady state Q^* and the stable periodic



Figure 3. (i) Bifurcation diagram showing stability of the steady state Q^* , along with the branch of periodic orbits which bifurcates from Q^* at the Hopf bifurcation points. All parameters except κ take values from Table 1. In all bifurcation diagrams stable (unstable) steady states are represented by solid (dotted) black lines, while the stable (unstable) limit cycles are represented by solid (dotted) blue lines. For periodic orbits we plot both the maximum and the minimum value of Q over the period, so the upper and lower curves both represent the same periodic orbit. Hopf bifurcation points *, saddle-node bifurcation of limit cycles points \Box , and period-doubling bifurcation points \bigcirc are indicated and also highlighted in insets. The saddle-node bifurcation in the right inset creates an interval for $\kappa \in (0.16872, 0.17632)$ of bistability between the periodic orbit and the steady state Q^* , while the pair of steady fold bifurcations seen in the left inset creates an interval for $\kappa \in (0.97254, 1.0247)$ of bistability between two different periodic orbits. (ii) Period of the periodic orbits seen in (i). Two period-doubling bifurcations and an interval for $\kappa \in (1.173, 1.2506)$ of period-doubled solutions are also shown. These can also be seen in the left inset in (i).

orbit of amplitude close to 1. Similar bistability has been observed before in hematopoiesis models. Bernard, Bélair, and Mackey [5, 6] studied a model for white blood cell (WBC) production which incorporated (1.1) to model the stem cell dynamics and found bistability for WBCs between a stable steady state and a stable periodic orbit. But in that model the bistability was seen by varying an amplification parameter in the WBC proliferation and was not associated with the variation of any parameter in the HSC equation. We are not aware of bistability having been observed previously in a standalone model for HSC dynamics.

There is a pair of fold bifurcations of periodic orbits which creates an interval for $\kappa \in (0.97254, 1.0247)$ of bistability of periodic orbits. Bistability is interesting, as it allows the possibility for a short-term perturbation of the system (such as during treatment) to cause the solution to switch between one solution and another and for the new stable dynamics to persist indefinitely. The two instances of bistability observed here occur for relatively small parameter intervals far from the homeostasis parameters and so are unlikely to be of great direct physiological relevance for healthy subjects. However, the existence of bistability is interesting in the context of dynamical diseases which are related to bifurcations that occur when parameters in the system are varied. There is also a pair of period-doubling bifurcations on the branch of periodic orbits illustrated in Figure 3. This leads to an interval for $\kappa \in (1.173, 1.2506)$ where a period-doubled orbit is stable.

The amplitude of the periodic orbits on the main branch tends to decrease along the branch, and the periodic orbits disappear in a supercritical Hopf bifurcation at $\kappa \approx 1.5317$. For κ between this value and its bound given by (2.8), the steady state Q^* is stable. The periods of the orbits shown in Figure 3(ii) strongly correlate with the amplitude of the orbits. The periods are all larger than a week, much larger than the delay $\tau = 2.8$ days.

Figure 4 illustrates some of the more interesting periodic orbits found during the κ continuation, including examples of bistability and period doubled orbits. The left panels show periodic solution profiles over one period, while the right panels display the time-delay embedding of the same solutions. Recalling that Q(t) is a scalar, but that the DDE (1.1) defines an infinite-dimensional dynamical system, $(Q(t), Q(t - \tau))$ gives a useful two-dimensional projection of the infinite-dimensional solutions, which has been used widely since it was introduced by Glass and Mackey [25]. Since it is only a projection of phase space, orbits may appear to cross each other, but because it incorporates both the terms Q(t) and $Q(t - \tau)$ that appear in (1.1) this projection is often very revealing.

Figure 5 shows the results of applying one-parameter continuation in the apoptosis rate γ , with the other parameters all held at their values in Table 1. The steady state Q^* is stable unless γ is close to its upper bound defined by (2.8). There is again a pair of Hopf bifurcations with the left bifurcation at $\gamma \approx 0.227918$ subcritical leading to an unstable periodic orbit with period ≈ 36.7 days at the bifurcation point. The period grows to about 82 days at a fold bifurcation of periodic orbits with $\gamma \approx 0.227766$. The periodic orbit becomes stable at the fold bifurcation leading to a very short interval of bistability between the stable periodic orbit and the stable steady state. As γ is increased from the fold bifurcation the stable periodic orbits gradually decrease in amplitude but increase in period reaching a maximum period of about 714 days when $\gamma \approx 0.2453692$. Some of these stable periodic orbits are illustrated in Figure 6. These orbits all have a single peak above Q^* which is only achieved once per period. After this peak the value of Q quickly drops to below 0.1, and there is then a very



Figure 4. Periodic orbits from the branch seen in Figure 3. (i) For $\kappa = 0.17$ the stable (solid line) and unstable (dotted line) periodic orbits which coexist with the stable steady state Q^* just before the subcritical Hopf bifurcation. (ii) The same periodic orbits shown in the two-dimensional $(Q(t), Q(t - \tau))$ projection of phase space. (iii)–(iv) For $\kappa = 1$ the two coexisting stable periodic orbits along with the unstable periodic orbit and unstable steady state. (v)–(vi) For $\kappa = 1.2$ the stable period-doubled orbit (solid line), along with the unstable periodic orbit from which it bifurcates (in panel (v) we show two periods of this orbit).

low amplitude oscillation in Q with a period of about 3 days (slightly larger than the delay $\tau = 2.8$) which decays in amplitude before the next spike in the number of HSCs. From (2.7), at $\gamma = 0.2453692$ the nonzero steady state is $Q^* = 0.089673$ and the longest period orbit shown in Figure 6 is close to homoclinic to the steady state Q^* ($Q'(t) \approx 0.00025$ when $Q(t) = Q^*$).

The long period orbits appear to be relaxation oscillations; these have been observed and studied previously for (1.1) [14, 24]. Visually, these solution profiles are more reminiscent of a spiking neuron [30] than what one would naively expect to see in blood cell concentrations. After the maximum period is achieved at $\gamma \approx 0.2453692$ the period declines precipitously to approximately 48 days at the Hopf bifurcation when $\gamma \approx 0.245375$ in an apparent canard



Figure 5. Continuation in γ with other parameters taking values from Table 1. (i) Bifurcation diagram showing stability of the steady state Q^* , along with the branch of periodic orbits which bifurcates from the steady state at the Hopf bifurcation points at $\gamma = 0.227918$ and $\gamma = 0.245375$. A saddle-node bifurcation seen in the top inset creates an interval for $\kappa \in (0.227766, 0.227918)$ of bistability between the periodic orbit and the steady state Q^* . (ii) Period of the periodic orbits seen in (i).



Figure 6. Three example periodic orbits from the branch shown in Figure 5. (i) Solution profiles and (ii) the corresponding $(Q(t), Q(t - \tau))$ delay embeddings, with the location of the orbits on the bifurcation branch indicated on the insets.

explosion. We are not aware of a canard being observed in a scalar DDE before, and we will investigate this phenomenon in section 4.

In Figure 7 we present the dynamics observed from applying one parameter continuation in the delay τ , with the other parameters held at their values in Table 1. The steady state Q^* is again stable unless the delay τ is close to its upper bound and is only unstable for τ between the pair of Hopf bifurcation points, which occur at $\tau = \tau_1^-$ and $\tau = \tau_1^+$, where τ^{\pm} are defined in (2.15). The left Hopf point is subcritical, leading to unstable orbits of period about 52 days, growing to a period of about 124 days at a saddle-node bifurcation of periodic orbits with $\tau \approx 5.72940$ days, where the periodic orbits become stable, creating an interval of bistability between the steady state and the stable periodic orbits. Ripples are visible in the amplitude of the branch of stable periodic orbits, with the magnitude of these undulations decreasing to zero as τ approaches the right Hopf bifurcation point, as shown in the bottom right inset of Figure 7(i). There are corresponding ripples in the period of the orbits, visible in the first inset of Figure 7(ii). The other insets show details of the branch of periodic orbits near the Hopf bifurcation points.



Figure 7. Continuation in τ with other parameters taking values from Table 1. (i) Bifurcation diagram showing stability of the steady state Q^* , along with the branch of periodic orbits which bifurcates from the steady state at the Hopf bifurcation points at $\tau = 5.74851$ and $\tau = 6.87437$. A saddle-node bifurcation seen in the top inset creates an interval for $\tau \in (5.72939, 5.74851)$ of bistability between the periodic orbit and the steady state Q^* . (ii) Period of the periodic orbits seen in (i).



Figure 8. Four stable periodic orbits from neighboring peaks and troughs of the ripples illustrate how they arise. As seen in the inset of the time-embedding plot, the large period and amplitude orbits from the peaks of the ripples perform one more small amplitude oscillation before escaping to $Q(t) \gg Q^*$ compared to the smaller period and amplitude solution from the preceding trough in the ripples.

Figure 8 illustrates stable periodic orbits from the left part of this branch. Although these orbits superficially resemble those of Figure 6, with a single peak above Q^* and a small amplitude oscillation close to Q = 0, the periodic orbits seen in Figure 8 have a quite different character to those seen in Figure 6. Specifically the orbits have a growing oscillation close to Q = 0 with a period close to the delay τ . In contrast, the solutions seen in Figure 6 have a decaying oscillation near their minimum value, which is not particularly close to Q = 0. Figure 8(ii) shows the delay embedding of the solutions, from which we see that the ripples in the amplitude and period along the branch are related to the number of low amplitude oscillations in the solution, with the smaller amplitude and period solutions at the bottom of the ripples performing one less oscillation in the $(Q(t), Q(t - \tau))$ projection before escaping to $Q(t) \gg Q^*$.

The dynamics of the oscillations close to Q(t) = 0 are easy to describe but harder to explain. The trivial steady state Q = 0 has one positive real characteristic value ($\lambda = 0.017605$

when $\tau = 6$) and infinitely many complex conjugate characteristic values, three pairs of which have positive real part. Thus the steady state Q = 0 of the DDE (1.1) has a seven-dimensional unstable manifold and an infinite-dimensional stable manifold in the full infinite-dimensional phase space of the functional differential equation. However, complex characteristic values would give rise to oscillatory solutions about Q = 0 which change sign. Since, by Theorem 2.1, solutions with positive initial conditions remain positive, oscillations about Q = 0 will not arise with physiological initial history functions. (Note that this implies that $|\omega| \ge \pi/\tau$, for any complex characteristic value $\lambda = \alpha \pm i\omega$, since otherwise τ would be less than half the period of the oscillation, and the positive half of the oscillation could be used to define an initial function $\varphi = -\varepsilon e^{\alpha t} \sin(\omega t)$ for $t \in [-\tau, 0]$ so the DDE that would have a solution close to $Q(t) = -\varepsilon e^{\alpha t} \sin(\omega t)$ for $0 < t \ll 1$, which would violate the positivity of solutions.) Consequently, in the restricted phase space of positive solutions that we consider Q = 0 has a one-dimensional unstable manifold and a trivial stable manifold.

In the inset of Figure 8(ii) we see that in the delay embedding $(Q(t), Q(t - \tau))$ that Q(t) takes its minimum value on the periodic orbit when $Q(t - \tau)$ is close to 0.5 but decreasing. Q(t) then increases slightly before decreasing again to its next minimum, which occurs very close to the minimum of $Q(t - \tau)$ on the solution. This sets in train a clockwise oscillation in the $(Q(t), Q(t - \tau))$ projection close to 0. Along this oscillation the local minima of Q(t) and $Q(t - \tau)$ occur very close to each other in time because the period of this low amplitude oscillation (as seen in Figure 8(i)) is very close to the delay τ . After the double local minima the solution grows close to the local unstable manifold of Q = 0 with $Q(t - \tau) \approx Q(t)e^{-\lambda\tau}$ for a time, until $Q(t - \tau)$ starts to decrease again (toward the previous local minima of Q(t)), after which Q'(t) becomes negative and Q(t) decreases to its next local minima, completing one cycle. The amplitude of this oscillation grows slightly with each subsequent cycle, until eventually the oscillation escapes to $Q(t) \gg Q^*$.

As τ increases across the branch of stable orbits the character of the periodic orbits changes (not illustrated), with the growth rate of the small amplitude oscillations progressively decreasing and the period of the orbit increasing. For τ sufficiently large the small amplitude oscillations decay instead of grow, and thereafter the periodic orbits resemble the longest period orbit shown in Figure 6(i). The period of the orbit, but not the amplitude, continues to grow until the amplitude and period of the solutions decrease abruptly just before the right bifurcation point, apparently in a canard explosion. The period reaches its maximum value of 3281 days for $\tau \approx 6.874295373$ and decreases dramatically to 182 days, while the value of τ remains constant to 10 significant digits (see right inset of Figure 7(ii)).

Although we do not find a homoclinic bifurcation, the longest period orbit is about 9 years, nearly 500 times larger than the delay in the system, with the orbit close to homoclinic to the nontrivial steady state Q^* . The solutions in this region are similar to the long period orbits displayed in Figure 6.

3.2. Two-parameter continuation. Recent versions of DDEBiftool [21, 53] have the facility to perform two-parameter continuation of bifurcations of periodic orbits, and we used this to study the bifurcations of the DDE (1.1) as the parameters κ , γ , and τ are varied pairwise.

In Figure 9 we present the two-parameter bifurcation diagram as κ and τ are varied, with all other parameters at their values in Table 1, which reveals the curves of Hopf,



Figure 9. Two parameter (κ, τ) -bifurcation diagram for the DDE (1.1) with other parameter values given by Table 1. Each solid curve represents the locus of bifurcation of periodic orbits, as indicated in the key. The brown square \Box denotes a Bautin bifurcation; see text. The black cross marks the homeostasis values of the parameters. Within the inset, the black square indicates the parameter values of the torus investigated in section 5.1, and the red and blue triangles represent chaotic attractors seen in section 5.2.

period-doubling, saddle-node, torus, and steady state bifurcation as this pair of parameters are varied. Taking a straight line through Figure 9 with $\tau = 2.8$ or with $\kappa = 0.022$ reveals the bifurcations found in Figures 3 and 7, respectively. From Figure 9 we see that the homeostasis parameters (κ, τ) = (0.022, 2.8) are not particularly close to any bifurcations, with the Hopf curve and an associated curve of saddle-node of limit-cycle bifurcations being the only other bifurcations near that part of parameter space. The Hopf bifurcations are subcritical to the left of the Bautin or generalized Hopf bifurcation at (κ, τ) = (0.4960, 1.525) and supercritical otherwise. We already saw instances of the subcritical Hopf bifurcations in Figures 3 and 7; Bernard, Bélair, and Mackey [6] previously presented an example with both Hopf bifurcations supercritical.

If the delay is small ($\tau < 1$) there are no bifurcations at all, while the bifurcation structures become more complicated as τ is increased with two curves of fold bifurcations of periodic orbits created in a cusp bifurcation at (κ, τ) \approx (0.81600, 2.3956), and a further cusp bifurcation and torus bifurcation curves only occurring for $\tau > 3$. Figure 9 suggests that for one-parameter continuation in κ , taking τ close to 4 will lead to more complicated dynamics than was seen in Figure 3 for $\tau = 2.8$. Indeed, the curves seen in Figure 9 were seeded by performing a oneparameter continuation in κ with $\tau = 3.9$ (see Figure 15) and consequently Figure 9 shows all the bifurcation curves that cross $\tau = 3.9$. There may be other bifurcation curves that remain above $\tau = 3.9$, but it appears from Figure 9 that they would be constrained to be near the right Hopf bifurcation.

Figure 10 shows the bifurcation curves found for two-parameter continuation in (κ, γ) with all the other parameters taking their values from Table 1, revealing an alternating sequence of curves of period-doubling and fold bifurcations (of limit cycles) and associated cusp bifurcation



Figure 10. Continuation in (κ, γ) with other parameters at homeostasis (see Table 1). Each solid curve represents the locus of a certain type of bifurcation of periodic orbits as specified in the key.

of limit cycles. The closed curves of bifurcations become shorter and narrower as parameters approach the upper bound on γ for periodic orbits to exist and also progressively more delicate to compute numerically. There may be additional curves of bifurcations for $\gamma > 0.2$ which we were not able to compute. A straight line through Figure 10 with $\gamma = 0.1$ or with $\kappa = 0.022$ reveals the bifurcations found in Figures 3 and 5, respectively. The inset reveals that the period-doubling and saddle-node loci do not overlap.

The results of two-parameter continuation in (γ, τ) are shown in Figure 11. This reveals the locus of the Hopf bifurcations already observed in Figures 5 and 7. There is also a Bautin bifurcation at $(\gamma, \tau) = (0.4020, 1.651)$ and a branch of saddle-node bifurcations of limit cycles which emerges this point. This branch represents the two-parameter continuation of the fold bifurcation seen in Figures 5 and 7. This bifurcation is very delicate for DDE-Biftool to compute and continue numerically, and we were not able to compute the full branch. For $\tau = 2.8$, 4, 6, and 8 we performed one parameter continuation in γ to confirm that the fold bifurcation persists for larger τ values and also to verify that there are not other bifurcation curves missing from the diagram. For the larger values of τ , DDEBiftool is not able to identify the fold bifurcation. While we are able to find the fold from a one-parameter continuation by simply looking for the minimum value of γ along the branch (and we added these points to Figure 11), DDEBiftool computes and continues fold bifurcations of limit cycles in two parameters by solving the defining equations for a fold bifurcation of periodic orbits [53], which is a considerably more complicated computation.

The two-parameter continuations in Figures 9, 10, and 11 reveal that the nontrivial steady state solution $Q^* > 0$ remains stable for all reasonably small perturbations from the homeostasis parameter values of Table 1. We also see from Figures 9 and 11 that Q^* is stable for all small delays $\tau < 1$ (at least when the other parameters are varied one at a time). This suggests that an ODE model would not capture the instabilities driven by the delays. Since the cell cycle time for stem cells is estimated to be much larger than 1 day (2.8 days in Craig,



Figure 11. Continuation in (γ, τ) with other parameters at homeostasis values. Each solid curve represents the locus of a bifurcation of periodic orbits, as specified in the key to Figure 10. The brown square \Box denotes a Bautin bifurcation. The brown dots and arc of brown curve emanating from the Bautin bifurcation form part of a curve of fold bifurcations of limit cycles, which DDEBiftool is only partially able to compute (see text).

Humphries, and Mackey [18]) it is essential to include the delay in the DDE model (1.1) to properly capture the possible dynamics of the system. That the two-parameter continuations in (κ, γ) and (γ, τ) reveal less interesting bifurcation diagrams than for continuation in (κ, τ) is probably not intrinsic to the properties of the parameters in the model but rather is determined by the homeostasis value of the third parameter from Table 1 when we perform two-parameter continuation. More complicated bifurcation diagrams can be generated by taking (κ, τ) close to (0.86, 4), in the interesting part of Figure 9, and then doing two-parameter continuation in any pair of these three parameters. For example, with $\kappa = 0.68$, two-parameter continuation in (γ, τ) (not shown) reveals torus and period-doubling curves, quite unlike anything seen in Figure 11. However, here we have based our continuation on using the homeostasis values of the parameters to vary it is likely that we could find more exotic dynamics, but what the relationship, if any, that dynamics would have to the Burns–Tannock HSC model is not clear.

4. Long period orbits and Canard explosion. A canard explosion is a dynamical phenomenon seen in fast-slow or singularly perturbed systems whereby over an exponentially small range of the continuation parameter a periodic orbit is transformed into a long period relaxation oscillation. For ODEs this requires at least two space dimensions, with classical examples being the van der Pol oscillator and Fitzhugh–Nagumo equations [4, 55]. Canard explosions have already been explored in DDEs [13, 34], but only in systems with at least two spatial dimensions that incorporate a delay. However, since DDEs are inherently infinitedimensional there is no reason why a canard explosion should not be seen in a scalar DDE such as (1.1).

DYNAMICS OF A HEMATOPOIETIC STEM-CELL MODEL

Recalling the continuation in γ shown in Figure 5, at the right Hopf bifurcation point $\gamma \approx 0.2453746$ a periodic orbit is born with period approximately 48 days, but at $\gamma \approx 0.2453692$ the period increases dramatically to about 700 days while the value of γ remains constant to seven significant figures. This would appear to be a canard explosion. Figure 12 illustrates orbits from this part of the branch as the period increases. Comparing the time plot with the $(Q(t), Q(t-\tau))$ phase space projection, the slow manifold appears to be close to $Q(t) = Q(t-\tau)$, with Q'(t) gradually increasing along this curve, followed by a fast transition layer as Q(t) decreases to close to its minimum value while $Q(t-\tau)$ remains close to its maximum. Then $Q(t-\tau)$ passes through the transition layer to also be close to its minimum value, after which there is a slowly decaying oscillation with a period of about 3 days as the solution converges back to the slow manifold. The largest period orbit illustrated has a period of about 701.3 days, with Q(t) crossing the steady state Q^* once in each direction, with $Q(t) > Q^*$ for approximately 191.3 days and $Q(t) < Q^*$ for the remaining 510.0 days. Here the delay $\tau = 2.8$ days, so this is an example of a (very) slowly oscillating periodic solution.

Fast-slow systems in ODEs often have separate fast and slow variables which can be considered separately in the fast and slow subsystems. That separation of variables does not occur in the DDE (1.1), for which we have only one variable. Nevertheless, relaxation oscillators with both fast and slow segments within the solution can arise and have been studied in DDEs, by



Figure 12. (i) Solution profiles and (ii) time-delay embedding $(Q(t), Q(t - \tau))$ of periodic solutions from the same γ continuation shown in Figure 5 show an apparent canard explosion at $\gamma \approx 0.2453692$. The insets indicate where the illustrated orbits lie on the branch. (iii)–(iv) Parts of (ii) at a much larger scale. (iii) Oscillatory convergence of the solutions onto a slow manifold for $Q < Q^*$. (iv) Dynamics near to Q^* . The nullcline Q'(t) = 0 is indicated by the red curves in panels (ii)–(iv).

tackling the fast and slow segments of the solution separately. In particular a relaxation oscillator for the HSC DDE (1.1) has been studied using singular perturbation analysis [14, 24]. Of particular note is the extensive work of Mallet-Paret and Nussbaum studying slowly oscillating periodic solutions in singularly perturbed constant and state-dependent DDEs [42, 43, 44].

A complete analysis of how the canard explosion arises in (1.1) will be beyond the scope of this work, but we will show that (1.1) can be considered as a singular perturbation problem. We identify the critical manifold and also investigate its persistence by approximating the resulting slow manifold and studying its stability. We will show that a segment of this manifold for $Q < Q^*$ is stable with oscillatory convergence of nearby trajectories onto the manifold (see Figures 12(ii) and (iii)), while a segment for $Q > Q^*$ is unstable, leading to the divergence of trajectories from the manifold. In the current work, we will not study the fast dynamics in the transition layer.

For $\gamma = 0.2453692$ and all other parameters taking their values from Table 1, we notice that $\kappa \approx 0$ and $A \approx 1$, so we introduce the perturbation parameter $\varepsilon \ge 0$ and let

(4.1)
$$\varepsilon = A - 1 = 2e^{-\gamma\tau} - 1, \qquad \kappa = \frac{\varepsilon f}{C}.$$

For the nonzero steady state Q^* to exist the inequality (2.8) must hold; equivalently the constant C must satisfy C > 1. Then $Q^* > 0$ is given from (2.7) by

(4.2)
$$Q^* = \theta (C-1)^{1/s},$$

independent of the value of $\varepsilon > 0$. For the parameters used in Figure 12 we have $\varepsilon = 6.132 \times 10^{-3}$ and C = 2.23 with $Q^* = 0.0896868$ when $\gamma = 0.2453692$. The parameter definitions in (4.1) could be applied to the nondimensionalized equation (2.3) with $\hat{A} = \varepsilon$ and $\hat{\kappa} = \varepsilon/C$, but we prefer to continue to study (1.1) directly.

Letting $h(Q) = Q\beta(Q)$, as in (2.10), and using (4.1) we rewrite (1.1) as

(4.3)
$$Q'(t) = -\frac{\varepsilon f}{C}Q(t) - h(Q(t)) + (1+\varepsilon)h(Q(t-\tau)).$$

When $\varepsilon = 0$ this reduces to

(4.4)
$$Q'(t) = -h(Q(t)) + h(Q(t-\tau)) = -\frac{fQ(t)}{1 + (Q(t)/\theta)^s} + \frac{fQ(t-\tau)}{1 + (Q(t-\tau)/\theta)^s}.$$

While (4.3) has the unique positive steady state Q^* given by (4.2), when $\varepsilon = 0$ there is a line of equilibria for (4.4) with Q being an arbitrary constant, which is the critical manifold. The linearization of (4.4) is given by (2.9) with a = -h'(Q) and b = -a and so the characteristic function (2.12) becomes

$$p(\lambda) = \lambda - a + ae^{-\lambda\tau}.$$

This satisfies p(0) = 0 and $p'(0) = 1 - a\tau$ and has $\lambda = 0$ as a solution for any value of a. There is an additional real negative root if p'(0) > 0, i.e., $a\tau < 1$. This root crosses zero when p'(0) = 0 and becomes positive for $a\tau > 1$ when p'(0) < 0. Thus the steady state stability changes when $h'(Q) = -1/\tau$. Using (2.1) and (2.10) and the nondimensionalized variables of (2.3) the identity $h'(Q) = -1/\tau$ reduces to a quadratic equation for \hat{Q}^s :

$$\hat{Q}^{2s} + (2 - (s - 1)\hat{f})\hat{Q}^s + 1 + \hat{f} = 0.$$

Solving this with parameters corresponding to Figure 12 we find that the stability on the critical manifold changes when $Q = \theta \hat{Q} = 0.0893174$, very close to the value Q^* .

The critical manifold should persist where it is transversally hyperbolic as a slow manifold following the theory of Fenichel [22]. However, that theory was developed for ODEs in multiple space dimensions with at least one fast and one slow variable. Likewise, the previous examples of canards in DDEs considered systems with two spatial dimensions with one fast and one slow variable [13, 34]. For the scalar DDE (1.1) there is not an obvious separation into fast and slow variables and it is not apparent how to proceed rigourously. Nevertheless, it is apparent from Figure 12 that there is a slow manifold, and in the remainder of this section we will show how to approximate the slow manifold and determine its stability.

The slow manifold on which $Q' \approx 0$ should be close to the nullcline Q'(t) = 0, which from (1.1) is given by

(4.5)
$$0 = -\frac{\kappa}{f}Q(t) - \frac{Q(t)}{1 + (Q(t)/\theta)^s} + 2e^{-\gamma\tau}\frac{Q(t-\tau)}{1 + (Q(t-\tau)/\theta)^s}$$

If one of Q(t) or $Q(t - \tau)$ is fixed, then for Q'(t) = 0 with s = 2 from (4.5) the value of the other one is defined by a cubic equation. The resulting nullcline is displayed as the two red curves in Figure 12(ii), which are seen to be disjoint in Figure 12(iv), which shows an expanded view near to (Q^*, Q^*) . Typically, canard explosions are seen close to a bifurcation of the intersections of the nullclines of the slow and fast variables. Here we do not have separate fast and slow variables, but we see that we are close to a bifurcation of the Q'(t) = 0nullcline itself, with the two disjoint parts coming very close to each other in Figure 12(iv). Figure 12(iv) also shows that the periodic orbits that form the canard explosion appear to lie on a slow manifold between the branches of the nullcline and switch from following the lower branch to following the upper branch at the point close to Q^* where the two curves are closest. It is thus likely essential for the canard explosion that the parameter set is close to this bifurcation of the nullcline structure.

To obtain a simple approximation to the slow manifold, let $h(Q) = Q\beta(Q)$, as in (2.10), so (1.1) becomes

(4.6)
$$Q'(t) = -\kappa Q(t) - h(Q(t)) + Ah(Q(t-\tau)).$$

Then use the approximation

(4.7)
$$h(Q(t-\tau)) \approx h(Q(t)) - \tau \frac{d}{dt} h(Q(t)) = h(Q(t)) - \tau Q'(t) h'(Q(t))$$

to remove the delay from (4.6). Note that $\tau \gg 0$ so (4.7) is useful only if $\tau^2 \frac{d^2}{dt^2} h(Q(t)) \ll 1$. Substituting (4.7) into (4.6) and rearranging we obtain

$$Q'(t) = \frac{-\kappa Q(t) + (A-1)h(Q(t))}{1 + A\tau h'(Q(t))}.$$

On the slow manifold $Q(t - \tau) \approx Q(t) - \tau Q'(t)$ so we can approximate the manifold in the delay embedding $(Q(t), Q(t - \tau))$ by the curve (Q, Q_{τ}) where



Figure 13. Approximations (Q, Q_{τ}) to the slow manifold of the canard solution where (i) Q_{τ} is given by (4.8) and shown in violet; (ii) Q_{τ} is given by (4.17) with Q_{τ} replaced by Q and shown in green. Both approximations are shown in the insets.

(4.8)
$$Q_{\tau} = \frac{(1 + \kappa \tau + A\tau h'(Q))Q - \tau (A - 1)h(Q))}{1 + A\tau h'(Q)}.$$

This gives a good approximation to the slow manifold for $Q \in (0, 0.23)$, except for a very small interval $|Q-Q^*| < 0.003$ about the nontrivial steady state. This is shown in Figure 13(i). The curve (Q, Q_τ) is also an approximation to the unstable manifold of Q = 0; the trivial steady state has a single positive characteristic value and infinitely many pairs of complex conjugate characteristic values with negative real part. The canard explosion is *not* associated with a solution homoclinic to Q = 0 though; no such homoclinic solution can exist in the space of nonnegative solutions, as all solutions in the stable manifold of Q = 0 will be oscillatory and violate the positivity of solutions (as already noted earlier).

To determine the dynamics close to the slow manifold, it is necessary to take proper account of the delayed term, and so we will derive another slow manifold approximation. For this, linearize h(Q) about $Q = Q_r$ for general Q_r as

(4.9)
$$h(Q) \approx h(Q_r) + h'(Q_r)(Q - Q_r).$$

Then let $Q_s(t)$ be a solution on the slow manifold, and let Q(t) be some other solution in a neighborhood of this manifold, and let $w(t) = Q(t) - Q_s(t)$ be the difference between these two solutions, then using (4.9) and (4.6) we find that

(4.10)
$$w'(t) = Q'(t) - Q'_s(t) \approx -(\kappa + h'(Q_s(t)))w(t) + Ah'(Q_s(t-\tau))w(t-\tau).$$

The linearization (4.10) is actually valid as an approximation of the dynamics about any solution $Q_s(t)$ of the DDE (1.1). However, it is problematic to use, even about solutions on the slow manifold, since although (4.10) is linear it is nonautonomous and the time-dependent terms depend on the as yet unknown slow-manifold solution $Q_s(t)$. As an alternative, rather than linearize about a particular solution, close to the slow manifold we can use (4.9) to linearize h about $Q = Q_r$ for some reference value of Q and convert (1.1) into a linear DDE for the dynamics near to $Q = Q_r$, with

(4.11)
$$Q'(t) \approx (A-1)[h(Q_r) - h'(Q_r)Q_r] - (\kappa + h'(Q_r))Q(t) + Ah'(Q_r)Q(t-\tau).$$

We rewrite (4.11) as

(4.12)
$$Q'(t) = aQ(t) + bQ(t - \tau) + c$$

where letting

(4.13)
$$G(Q) := (A-1)\beta(Q)Q - \kappa Q = (A-1)h(Q) - \kappa Q$$

we see that the constants a, b, and c satisfy

$$a = -(\kappa + h'(Q_r)), \qquad b = Ah'(Q_r),$$

$$c = (A-1)[h(Q_r) - h'(Q_r)Q_r] = G(Q_r) + [\kappa - (A-1)h'(Q_r)]Q_r = G(Q_r) - (a+b)Q_r.$$

Noting that

$$G'(Q_r) = (A-1)h'(Q_r) - \kappa = a+b,$$

we can rewrite c as $c = G(Q_r) - G'(Q_r)Q_r$.

The function G is unimodal, and recalling (2.5) it follows that $G(0) = G(Q^*) = 0$, and G has a unique maximum for $Q = Q_f$ for some $Q_f \in (0, Q^*)$. Thus $G'(Q_r) > 0$ (resp., $G'(Q_r) < 0$) for $Q < Q_f$ (resp., $Q > Q_f$). Another value of Q which will be relevant below is Q_h , the value of Q such that $h'(Q_h) = 0$. It follows easily that $Q_h > Q_f$. With $\gamma = 0.2453692$ and the other parameters from Table 1 we have

$$Q_f = 0.042263 \quad < \quad Q_h = (s-1)^{-\frac{1}{s}}\theta = \theta \quad < \quad Q^* = 0.089686.$$

Solutions to the nonhomogeneous linear DDE (4.12) consist of a particular solution of (4.12) and any linear combination of solutions of the homogeneous linear DDE

(4.14)
$$Q'(t) = aQ(t) + bQ(t - \tau).$$

The DDE (4.14) is of the same form as (2.9) and has the same characteristic equation (2.12). This has infinitely many complex roots, which would lead to oscillatory solutions of (4.14). However, from above the slow manifold appears to be monotonic, hence we will seek a monotonic solution of (4.11), for which we require real roots of (2.12). Since $p'(\lambda) = 1 + b\tau e^{-\lambda\tau}$, for $Q \leq Q_h$ we have $p'(\lambda) \geq 1$ and (2.12) has a unique real root. Real roots of (2.12) can be found using the Lambert-W function [17]. Rearranging (2.12) we see that any root λ satisfies $b\tau e^{-a\tau} = (\lambda - a)\tau e^{(\lambda - a)\tau}$. Hence $W(b\tau e^{-a\tau}) = (\lambda - a)\tau$, and so

(4.15)
$$\lambda = a + \frac{1}{\tau} W(b\tau e^{-a\tau}).$$

Consider first the case where $Q_r < Q_f$. Then $G'(Q_r) = a + b > 0$ and b > 0 > a. Since p(0) = -(a + b) and $p'(\lambda) \ge 1$ the characteristic equation (2.12) has a unique real root $\lambda_+ = \lambda_+(Q_r) \in (0, G'(Q_r))$ given by (4.15). Equation (4.12) then admits a constant solution, Q(t) = k, where

$$k = \frac{-c}{a+b} = \frac{-G(Q_r) + G'(Q_r)Q_r}{G'(Q_r)} = Q_r - \frac{G(Q_r)}{G'(Q_r)}.$$

Hence a monotonic solution of (4.12) passing through $Q(0) = Q_r$ is

(4.16)
$$Q(t) = Q_r + (e^{\lambda_+ t} - 1)G(Q_r)/G'(Q_r).$$

The general solution of (4.12) which describes the behavior of solutions in a neighborhood of the monotonic solution is obtained by adding an arbitrary linear combination of the solutions of (4.14), defined by the roots of (2.12). But since $\lambda_+ > 0$, even without solving for the complex characteristic values, we already know that the monotonic solution is not stable for $Q < Q_f$. Nevertheless, we can use (4.16) to approximate the slow manifold for $Q < Q_f$. From (4.16), when $Q = Q_r$ we have $Q' = \lambda_+ G(Q_r)/G'(Q_r) < G(Q_r)$, and hence assuming that $Q(t - \tau) \approx Q(t) - \tau Q'(t)$, we can approximate the slow manifold $Q_s(t)$ in the delay embedding $(Q(t), Q(t - \tau))$ for $Q < Q_f$ by the curve (Q_r, Q_τ) where

$$(4.17) Q_{\tau} = Q_r - \tau \lambda G(Q_r) / G'(Q_r)$$

and $\lambda = \lambda_+(Q_r)$.

Next consider the case for which $Q_r \in (Q_f, Q_h)$. Then $b > 0 > a + b = G'(Q_r)$. Now, p(0) = -(a+b) > 0 and $p'(\lambda) \ge 1$, so the characteristic equation (2.12) has a unique real root which is negative, $\lambda_- = \lambda_-(Q_r) \in (G'(Q_r), 0)$ given by (4.15). Similarly to above, (4.12) then has a monotonic solution

(4.18)
$$Q(t) = Q_r + (e^{\lambda_- t} - 1)G(Q_r)/G'(Q_r),$$

passing through $Q(0) = Q_r \in (Q_f, Q_h)$. The behavior of nearby solutions is determined by the general solution of (4.12), which is

(4.19)
$$Q(t) = Q_r + (e^{\lambda_- t} - 1) \frac{G(Q_r)}{G'(Q_r)} + \beta_0 e^{\lambda_- t} + \sum_{j \ge 1} e^{\alpha_j t} (\beta_j \cos(\omega_j t) + \gamma_j \sin(\omega_j t))$$

for constants β_j , γ_j where $\lambda_j = \alpha_j \pm i\omega_j$ are the complex roots of (2.12). For this solution, not only is $\lambda_- < 0$, but we can also show that all the complex characteristic values that solve (2.12) also have strictly negative real part. Taking real and imaginary parts of (2.12) we find that $\lambda_j = \alpha_j \pm i\omega_j$ satisfies $0 = \alpha_j - a - be^{-\alpha_j\tau} \cos(\omega_j\tau) = \omega_j + be^{-\alpha_j\tau} \sin(\omega_j\tau)$, which implies that $\omega_j^2 = b^2 e^{-2\alpha_j\tau} - (\alpha_j - a)^2$. But for $Q_r \in (Q_f, Q_h)$ we have a < 0 and -a > b > 0, hence for a characteristic root with $\alpha_j \ge 0$ we have $\omega_j^2 \le b^2 - a^2 < 0$, a contradiction, and so all characteristic values have $\operatorname{Re}(\lambda_j) = \alpha_j < 0$.

For $Q \in (Q_f, Q_h)$ we can again approximate the slow manifold $Q_s(t)$ in the delay embedding $(Q(t), Q(t - \tau))$ by the curve (Q_r, Q_τ) where Q_τ is defined by (4.17) with $\lambda = \lambda_-(Q_r)$. Since $\alpha_j < 0$ for all j and $\lambda_- < 0$ all the additional solution elements included in (4.19) are decaying, and the solution defined by (4.18) and the resulting slow manifold are attracting in this region of phase space.

The convergence onto the slow manifold is oscillatory, as seen in Figure 12. This is governed by the dominant complex characteristic value of (2.12), the value of which is stated as λ_1 in Table 2. We see from the table that $\operatorname{Re}(\lambda_1)$ becomes more negative as Q increases, implying that the slow manifold becomes more attractive as Q increases toward Q_h . This is clearly visible in Figure 12(iii) with progressively fewer oscillations visible for the orbits converging onto the slow manifold for larger values of Q. The period of these oscillations $2\pi/\omega_1$ also increases with Q, but not greatly, and is close to 3 in the range of Q values where the oscillations are most visible.

Table 2

The real characteristic value λ_{-} or λ_{+} of (2.12) along with the first pair of complex characteristic values λ_{1} , the derivative of the solution Q'(t) given by (4.16) or (4.18), and the resulting approximation of Q_{τ} given by (4.17) for a range of values of Q_{r} .

Q_r	λ_{\pm}	Q'	$Q_{ au}$	λ_1
0.01	1.10×10^{-5}	1.17×10^{-5}	9.96×10^{-3}	$-0.0118 \pm 2.15i$
0.02	9.56×10^{-4}	2.45×10^{-5}	1.99×10^{-2}	$-0.0155 \pm 2.13i$
0.03	6.68×10^{-4}	3.96×10^{-5}	2.98×10^{-2}	$-0.0237 \pm 2.11i$
0.04	1.60×10^{-4}	5.81×10^{-5}	3.98×10^{-2}	$-0.0408 \pm 2.07i$
0.05	-7.40×10^{-4}	8.13×10^{-5}	4.97×10^{-2}	$-0.077\pm2.00i$
0.06	-2.45×10^{-3}	1.10×10^{-4}	5.96×10^{-2}	$-0.157\pm1.90i$
0.07	-6.28×10^{-3}	1.47×10^{-4}	6.95×10^{-2}	$-0.354\pm1.75i$
0.08	-1.93×10^{-2}	1.98×10^{-4}	7.94×10^{-2}	$-1.35 \pm 1.42i$



Figure 14. (i) Profile and (ii) time-delay embedding, for a periodic solution of the nonlinear DDE (1.1), and several of its approximations. For $\gamma = 0.2453692$ the DDE (1.1) has a limit cycle with period 297 days, part of which is shown here (blue curve). For the approximations we take $Q_r = 0.063224$; then the black curve shows the approximation (4.18) to the slow manifold. The red curve shows a solution to (4.12) computed numerically using a segment of the solution of the nonlinear DDE (1.1) for $t \in [-2.8, 0]$ to define the initial function. The green curve shows a solution of (4.12) defined by (4.19) with $\beta_j = \gamma_j = 0$ for all j, except $\gamma_1 = 0.003935$.

Figure 14 illustrates how well our approximations perform in the region $Q \in (Q_f, Q_h)$ where the slow manifold is attracting. The blue curves in Figure 14 show part of the limit cycle of the nonlinear DDE (1.1) with period 297 days when $\gamma = 0.2453692$, which occurs in the canard explosion and was previously shown in Figure 12. Taking $Q_r = 0.063224$ we find that the rightmost characteristic value is $\lambda_- = -3.33 \times 10^{-3}$; then (4.18) defines an approximation to the slow manifold which is shown as the black curve in Figure 14.

The second-rightmost characteristic value $\lambda_1 = \alpha_1 + i\omega_1 = -0.202 + 1.86i$ yields the approximate oscillation time of 3.37 days. To show that this characteristic value governs the convergence of solutions onto the slow manifold in Figure 14 we show as the green curve the solution (4.19) of the linearized DDE (4.12) with $\beta_j = \gamma_j = 0$ for all j, except for $\gamma_1 \neq 0$, so that the only oscillatory mode included in the solution is defined by λ_1 . Additionally, the red curve shows the solution of (4.12) incorporating all modes, computed by solving (4.12) numerically using part of the solution of (1.1) as the initial function. Both approximations have oscillations about the slow manifold with very similar period and decay rate as for the solution of the full nonlinear DDE (1.1), demonstrating the validity of our approximations. In the current work we will not describe the passage of the slow manifold past the steady state Q^* , but note that the behavior of the solutions of (2.12) changes when Q_r approaches Q^* . For $Q > Q_h$ we have b < 0 and the different branches of W(x) in (4.15) for x < 0 can lead to zero or two real solutions for λ . There are two values of Q, $Q_{h'}^- < Q^* < Q_{h'}^+$, such that when $Q = Q_{h'}^{\pm}$, we have $b\tau e^{-a\tau} = -e^{-1}$ and the two branches of the Lambert-W function coalesce. These points can be computed from the solution on each branch of $h'(Q_{h'}^{\pm}) = W(-e^{-1-\kappa\tau}/A)/\tau$, which leads to $Q_{h'}^- = 0.08626$ and $Q_{h'}^+ = 0.09389$. For $Q \in (Q_{h'}^-, Q_{h'}^+)$, (2.12) has no real roots. At the boundaries, $Q_{h'}^{\pm}$ or $Q > Q_{h'}^+$ there are two real characteristic roots coalesce, and for $Q < Q_{h'}^-$ or $Q > Q_{h'}^+$ there are two real characteristic roots with positive real part and leading characteristic roots of (2.12) are $\lambda_1 = 0.0070 \pm 0.1303i$ and $\lambda_2 = -0.73 \pm 2.67i$.

For $Q > Q_{h'}^+$, the function $p(\lambda)$ in (2.12) is convex with p(0) = -(a+b) > 0 and $p(\lambda) > 0$ for $\lambda \ge a$. With the other parameters as stated, (2.12) has two positive solutions provided $Q_r \le 0.28577$. Using the smaller of these two roots, we obtain a monotonic solution of the same form as (4.16) and (4.18), which can be similarly used to construct an approximation to the slow manifold for $Q > Q_{h'}^+$, as shown in Figure 13(ii). Because of the two positive characteristic roots, this part of the slow manifold is unstable, as seen in the dynamics where the periodic orbits of different amplitudes and periods are seen in Figure 12(ii) to peel away from each other sooner or later depending on their amplitude and period. Thus we have approximated the attracting and repelling parts of the slow manifold either side of the steady state Q^* . A complete analysis of the canard explosion would require the dynamics that join these segments of the slow manifold, both near to the steady state, and also the fast dynamics when the solution is far from the slow manifold.

5. Nonperiodic and chaotic dynamics. The period-doubling and torus (Neimark–Sacker) bifurcations that we found in section 3 suggest that the DDE (1.1) can generate quasi-periodic and chaotic dynamics. The software package DDEBiftool cannot be used to directly find such dynamics, but we can use the DDEBiftool bifurcation studies of section 3 to determine parameter regions where nonperiodic dynamics should arise. Long time numerical simulations of the DDE can then be performed using the MATLAB dde23 routine [45] with suitable initial history functions to study the dynamics in these parameter regimes.

5.1. Quasi-periodic dynamics. It is somewhat surprising to find torus bifurcations for the HSC model (1.1), because as we already noted in section 2, its linearization about a steady state is (2.9), which does not admit any double-Hopf bifurcations. Double-Hopf or Hopf-Hopf bifurcations are a standard mechanism for generating tori and curves of torus bifurcations [35] and arise frequently in systems with coupled oscillators and systems with multiple delays [12]. However, the HSC model (1.1) is scalar, with a single delay, but nevertheless torus bifurcations do occur, as seen in section 3.2. So here we will investigate the existence of invariant tori for (1.1).

The two-parameter continuation in (κ, τ) shown in Figure 9 reveals an isola of torus bifurcations for $\kappa \in (0.91859, 1.0174)$ and $\tau \in (3.4857, 4.1342)$. In a one-parameter continuation, as κ is varied with $\tau = 3.9$ fixed, Figure 15 reveals that the main branch of periodic orbits loses stability for $\kappa \in (0.95004, 0.97309)$ at a pair of torus or Neimark–Sacker bifurcations,



Figure 15. (i) Bifurcation diagram for one-parameter continuation in κ with $\tau = 3.9$ and other parameters taking homeostasis values from Table 1. Hopf bifurcation points *, saddle-node bifurcation of limit cycles points \Box , period-doubling bifurcation points \bigcirc , and torus bifurcation points \diamondsuit are indicated and highlighted in insets. (ii) Period of the limit cycles displayed in (i).

corresponding to the points where the continuation crosses the isola found in the twoparameter continuation.

The simplest explanation is that there should be a stable torus at κ values between the two Neimark–Sacker bifurcations. Although DDEBiftool [53] cannot be used to find tori directly, a stable torus can be found by direct numerical simulation if a suitable initial function is chosen in the basin of attraction of the torus. Some care needs to be taken, because as the top right inset in Figure 15(i) reveals, folds on the main branch of periodic solutions result in a stable (as well as two unstable) periodic orbits existing for $\kappa \in (0.088007, 1.1556)$, so if a stable torus exists it will coexist with a stable periodic orbit.

To confirm the existence of a stable torus we performed a long time integration of the DDE using the MATLAB dde23 routine [45] with initial history function very close to the unstable periodic orbit on the main branch of solutions. For $\kappa = 0.961$ with $\tau = 3.9$ and all other parameters taking their values from Table 1 (this parameter combination is indicated by the black square in the inset within Figure 9) we found a quasi-periodic torus which envelops the unstable periodic orbit, as illustrated in Figure 16. The existence of the quasi-periodic torus was confirmed numerically both by plotting the Poincaré section and by computing the Lyapunov exponents.

Recall from section 2 that the DDE (1.1) has the infinite-dimensional phase space $C = C([-\tau, 0], \mathbb{R})$; consequently a hyperplane defined by a Poincaré section is also infinitedimensional. For $\alpha \in [0, \tau]$ and some constant $c \in \mathbb{R}$ we define the Poincaré section $\mathcal{P}_{\alpha} := \{u_t \in C : u_t(-\alpha) = c, u'_t(-\alpha) > 0\}$. For $\alpha = 0$ this is equivalent to looking for the points \hat{t} along the solution trajectory such that $u(\hat{t}) = c$ and $u'(\hat{t}) > 0$ and taking as the corresponding element of the Poincaré section the function segment u(t) for $t \in [\hat{t} - \tau, \hat{t}]$, so that u is equal to c at the right-hand end of the function segment. Other choices of α are also possible, so, for example, with $\alpha = \tau$ the function u will be equal to c at the left-hand end of the interval.

For the Poincaré section \mathcal{P}_0 to be useful we need to project it into finite dimensions. The simplest way to do this is to take the value of the solution u(t) at a finite set of points in $[t - \tau, t]$. Since the choice $\alpha = 0$ fixes u(t) = c, we choose the time points $t - \tau/2$ and



Figure 16. For $\tau = 3.9$ and $\kappa = 0.961$. (i) Time series and (ii) projected solution space $(Q(t), Q(t - \tau))$ for a quasi-periodic orbit (red curve) and the three periodic solutions (blue curves) seen in the top right inset of Figure 15(i) for $\kappa = 0.961$. (iii) Projected Poincaré section of the quasi-periodic orbit (red) and the unstable orbit (blue) that it envelops onto the plane $(Q(t - \tau), Q(t - \tau/2))$ for crossing of the Poincaré section Q(t) = c = 0.14with Q'(t) > 0. (iv) Initial convergence of the first three Lyapunov exponents.

 $t - \tau$ and project the Poincaré section into \mathbb{R}^2 by plotting $u(t - \tau/2)$ against $u(t - \tau)$ for values of t such that u(t) = c. This is equivalent to the projection $P : \mathcal{P}_0 \to \mathbb{R}^2$ defined by $P(u_t) = (u(t - \tau), u(t - \tau/2))$. Figure 16(iii) reveals the results of doing this with c = 0.14 for both the putative torus and the unstable periodic orbit that gave rise to it. This reveals the expected torus structure with the points representing the function segments in \mathcal{P}_0 lying on a closed curve that encloses the point representing the periodic orbit in the two-dimensional projection. Since each of the red points represents separate intersections of the same orbit with the Poincaré section, the orbit is either quasi-periodic or of period longer than a human adult lifespan (the time integration was 30000 days, which is longer than 82 years).

We computed the Lyapunov exponents of the quasi-periodic orbit on the torus using the method of Breda and Van Vleck [9]. Figure 16(iv) shows the initial convergence of the numerical estimates for the three largest Lyapunov exponents. After 3×10^4 days, the six largest Lyapunov exponents are estimated to be 0.00052, -0.00066, -0.0093, -0.18, -0.29, and -0.29. This reveals that up to the numerical accuracy the first two exponents are both zero, and the rest are negative, as is characteristic for a quasi-periodic two-torus.

As is well known in torus dynamics [10], perturbing parameters in the system will change the dynamics on the torus, with parameter regions of phase locking, where there is a stable periodic orbit on the torus, interspersed with parameter sets for which the dynamics are truly quasi-periodic. So, although we cannot prove that there exists a quasi-periodic torus for



Figure 17. Time series over one period and the corresponding time-delay embedding $(Q(t), Q(t - \tau))$ for stable periodic orbits on the phase-locked torus with $\tau = 3.9$ for (i)–(ii) $\kappa = 0.965$, and (iii)–(iv) $\kappa = 0.957$, which intersect the Poincaré section \mathcal{P}_0 three and seven times, respectively.

exactly the parameters illustrated in Figure 16, there will be for nearby parameter values. Equally, there will be parameter sets for which phase locking occurs on the torus, leading to stable periodic orbits of large period. In Figure 17 we show examples with $\kappa = 0.957$ and $\kappa = 0.965$ and all the other parameters at their values for the example of Figure 16 (in the (κ, τ) space of Figure 9 both these parameter sets are inside the torus curve close to the black square). These show stable periodic orbits which close after going around the torus 7 and 3 times, leading to periodic orbits of periods approximately 90.5 and 38.3 days. These orbits intersect the Poincaré section \mathcal{P}_0 and its projection into \mathbb{R}^2 seven and three times, respectively.

The phase-locked orbits exist over a parameter region called an Arnold tongue. DDEBiftool can be used to find the edges of these Arnold tongues (which are bounded by a fold bifurcation of periodic orbits between the interleaved stable and unstable orbits that lie on the torus inside the parameter region of the Arnold tongue). These Arnold tongues will lie in the small parameter region indicated in the inset of Figure 9 where torus bifurcations occur. We will not pursue the Arnold tongue structure in this work; Arnold tongues have previously been computed for DDEs, even in the state-dependent delay case [12].

Another curve of torus bifurcations is visible in Figure 9 close to $(\kappa, \tau) = (0.7, 4)$. The corresponding torus bifurcation can be seen in Figure 15 at $\kappa \approx 0.65046$ (with $\tau = 3.9$), where the periodic orbit loses stability in a torus bifurcation. In this case there is not a second corresponding torus bifurcation where the periodic orbit regains stability. Instead there is a period-doubling bifurcation (which is a resonant torus bifurcation) near $\kappa = 0.767$, but the periodic orbit on the principal branch does not regain stability at this point. That this

period-doubling bifurcation is associated with the neighboring torus bifurcation can be surmised from Figure 9, where we see that the endpoints of the curve of torus bifurcations lie on the period-doubling bifurcation curve. The torus dynamics are likely to be more complicated in this case, but we did not explore them.

5.2. Chaotic dynamics. Having found long period and quasi-periodic orbits, it is natural to also ask whether (1.1) admits chaotic solutions. Kaplan and Yorke [31] defined an attractor dimension, now known as the Lyapunov dimension, to be

(5.1)
$$d = k - \frac{1}{\lambda_{k+1}} \sum_{j=1}^{k} \lambda_j,$$

where the Lyapunov exponents are ordered so $\lambda_1 \ge \lambda_2 \ge \ldots$, and k is the largest integer so that the sum of the first k exponents is nonnegative, thus necessarily $\lambda_{k+1} < 0$, and $d \in [k, k+1)$. For the torus seen in the previous section with $\lambda_1 = \lambda_2 = 0 > \lambda_3$, the dimension given by (5.1) is d = 2, as expected for a torus.

One generally accepted indication of chaos is the presence of a positive Lyapunov exponent, in which case the Lyapunov dimension will be larger than two. We will investigate the existence of chaotic solutions for (1.1) by numerically computing the Lyapunov exponents, again using the method of Breda and Van Vleck [9].

In Figure 15 we see that for $\kappa \in (0.795, 0.929)$ both steady states and the periodic orbits on the main branch and on the period doubled branch are all unstable. However, from Theorem 2.1 we know that the dynamics must remain bounded, and so there must be a global attractor for these parameters. This parameter interval of unstable solutions for the κ continuation lies between two period doubling bifurcations in Figure 15, which is also inside the lobe of period-doubling bifurcations depicted in the two-parameter continuation in κ and τ in Figure 9, and we investigate the dynamics within this region.

In Figure 18 we present an orbit diagram for (1.1) as τ is varied across this region with $\kappa = 0.865$. Orbit diagrams are usually produced for maps, and we reduce the solution of (1.1) to a map by considering the crossings of a Poincaré section. Previously, we considered Poincaré sections with Q(t) constant, which would not work so well in this case because the value of Q^* changes as τ is varied, and we would need to vary the constant to ensure that the orbits cross the Poincaré section. Instead, we consider the local maxima and minima of Q(t)along the solution, or equivalently the points where Q'(t) = 0 with Q''(t) < 0 or Q''(t) > 0(respectively). For each value of τ using the MATLAB dde23 routine [45] we integrate through a time interval of 50τ days, then plot the value of Q at its last local maxima and minima. Since the dynamics are more interesting for some τ values than others, we defined a τ mesh with 9121 points from 1 to 3.4, 19000 points from 3.4 to 4.4, and 2281 points from 4.4 to 5. These three meshes were combined to form a mesh of 30400 points from 1 to 4 for increasing τ . A second mesh with 30399 points interleaved with the previous mesh was used for decreasing τ . For each mesh point the last τ time interval of the solution was used as the initial function to compute the solution at the next mesh point. The results displayed in Figure 18 clearly reveal the bifurcations already shown in Figure 9 including the Hopf bifurcations at $\tau = 1.1364$ and 4.6841, the fold bifurcations near $\tau = 2.4379$, 2.4451, 4.3281, and 4.4364, and the period-doubling bifurcations at $\tau = 3.1303, 4.3909, 4.4215, \text{ and } 4.5575$. Between



Figure 18. Orbit diagram showing local maxima and minima of solutions of (1.1) as a function of the delay τ , with $\kappa = 0.865$ and other parameters taking their values from Table 1. The red and green dots denote respectively the local maxima and minima computed along a mesh with 30400 points for increasing τ , while the black and blue dots denote the local maxima and minima computed by decreasing τ . The bifurcation points of primary branch are identified using the same symbols as in Figures 3 and 15. The upper side panel shows a detail from the main panel, while the other two side panels show just the decreasing and increasing parameter scans, illustrating bistability and hysteresis in the system.



Figure 19. A sequence of windows of periodic dynamics with parameter intervals of apparent chaotic dynamics. Panel (i) shows a zoom of part of the Figure 18, while panel (ii) shows a zoom of part of panel (i).

those period-doubling bifurcations, Figure 18 reveals numerous period-doubling cascades and several parameter intervals of apparent chaotic dynamics with windows of periodic dynamics.

For some intervals of parameter values the results of sweeping left to right and right to left are significantly different, revealing the bistability of attracting states and hysteresis between them. The side panels to Figure 18 illustrate this for $\tau \approx 3.85$, where increasing τ sequentially appears to reveal chaotic dynamics, but decreasing τ reveals a stable periodic orbit which appears to undergo a period-doubling cascade leading to a small interval of parameter values for $\tau \approx 3.9$ for which there are apparently coexisting chaotic attractors.

Figure 19 shows two successive magnifications from a small region of Figure 18. To reveal the finer structure, we recomputed the orbit diagram for each of these τ intervals for 30000

equally spaced increasing τ values, and a second interleaved mesh with one fewer point with decreasing τ values. Figures 18 and 19(i)–(ii) together suggest a self-similarity of the structure with sequences of windows of periodic dynamics separated by intervals of apparent chaotic dynamics, on ever smaller parameter intervals. While it would be interesting to study the scaling in the period doubling cascades, this is very difficult to do because the mapping is only implicitly defined and requires that we numerically solve the DDE (1.1) between each extremum of Q(t). Instead here, we will investigate the nature of the chaotic solutions.

With $\kappa = 0.865$ and $\tau = 3.9$ the orbit diagram suggests that the dynamics should be chaotic, and this case is illustrated in Figure 20. At first glance the time series in panel (i) resembles a period-doubled solution, but the maxima close to Q(t) = 0.2 actually alternate in height, so the solution is closer to a period-quadrupled solution. However, the time-delay embeddings in panels (ii) and (iii) appear to show that the orbit is not periodic but that there is a very structured low-dimensional attractor. The solutions were computed by taking a constant initial history function close to Q^* and integrating with dde23 through the transient dynamics until the orbit converges to the attractor. The segment of the solution trajectory that is displayed in Figures 20(ii) and (iii) spans 3000 days. The initial convergence of the first three Lyapunov exponents is illustrated in Figure 20(iv), but the full computation of the exponents, using the method of Breda and Van Vleck [9], is over a time interval of 30000 days, or 82 years. The dynamics are not periodic over this time interval and the leading Lyapunov exponents are



Figure 20. Chaotic orbit for $\tau = 3.9$ and $\kappa = 0.865$. (This parameter set is indicated by the red triangle in Figure 9 inset.) (i) Orbit segment. (ii) Time-delay embedding $(Q(t), Q(t - \tau))$ and (iii) solution space $(Q(t), Q(t - \tau/2), Q(t - \tau))$ of the solution for $t \in [3000, 6000]$. The black line and black dots in (i)–(ii) represent the unstable steady state Q^* . (iv) Convergence of the first three Lyapunov exponents.

computed numerically to be 0.0107, -0.0002, -0.0966, and -0.1577. The second Lyapunov exponent here is 0 to numerical accuracy, and the presence of a positive Lyapunov exponent indicates chaos. The appearance of the orbit being close to a period-quadrupled orbit is most likely just due to the provenance of the chaotic orbit being created through a period-doubling cascade. Had we only looked at the time series we could have been wrongly led to conclude that the dynamics was not chaotic; the time series of the solution alone is very rarely sufficient to determine the nature of the dynamics in the interesting cases. For the attractor shown in Figure 20 the Lyapunov dimension is computed from (5.1) to be d = 2.11.

As can be seen from Figure 18 the character of the chaotic dynamics is very sensitive to changes in the parameter values. Changing τ from 3.9 to 4.07 while keeping κ and all the other parameters at their values in Figure 20 (see the blue triangle in Figure 9 inset) the dynamics becomes as shown in Figure 21. Now the time series in Figure 20(i) is visually nonperiodic, and the time-delay embedding in Figures 20(ii) and (iii) appears to fill more of the phase space. This is reflected in the Lyapunov dimension. Computing out to 1.8×10^5 days (about 500 years) the leading Lyapunov exponents are estimated to be 0.03027, -0.00009, -0.11271, and -0.153236. Using (5.1) the Lyapunov dimension is computed to be d = 2.268, larger than in the previous example.

The Lyapunov exponents could have been obtained with a shorter integration interval; the reason to integrate out to 500 years was to obtain many crossings of the Poincaré section



Figure 21. Chaotic orbit for $(\kappa, \tau) = (0.865, 4.07)$. (i) Segment of solution time series on the attractor. Time-delay embeddings: (ii) $(Q(t), Q(t-\tau))$, and (iii) $(Q(t), Q(t-\tau/2), Q(t-\tau))$ for $t \in [3000, 6000]$. The black line and black dots in (i)–(iii) represent the unstable steady state. (iv) Projection $P(u_t) = (u_t(-\tau/2), u_t(-\tau))$ of function elements u_t in the Poincaré section $\mathcal{P}_{\tau} = \{u_t : u_t(0) = Q^*, u'_t(0) < 0\}$.

 $Q(t) = Q^*$ in order to try to reveal the fractal structure of the attractor. This is difficult to achieve because the mapping between the intersections with the Poincaré section is only implicitly defined by the solution of the DDE (1.1) which has to be solved numerically. Nevertheless Figure 21(iv) shows a projection of the crossing of the Poincaré section, with insets which reveal some of the fractal structure of the attractor.

If we vary all three of κ , γ , and τ , while still holding all the other parameters at their homeostasis values from Table 1, further interesting chaotic solutions can be found. Figure 22 shows an orbit that appears to display transient chaos. We interpret this as coexistence of a chaotic invariant set which is not asymptotically stable along with a periodic orbit which is stable. The orbit initially appears to be chaotic with a high-dimensional attractor (see panels (ii) and (iii)) but after about 2850 days transitions to the stable period-doubled periodic orbit which has a period of about 87.75 days. This orbit was found by taking parameters close to a point where two period-doubling bifurcation branches cross each other in a bifurcation



Figure 22. For $\kappa = 0.68$, $\gamma = 0.0354608$, and $\tau = 9.88888$ an orbit which appears to display transient chaos. (i) The transition from nonperiodic to periodic motion. (ii)–(iii) Time series Q(t) and delay embedding $(Q(t - \tau), Q(t))$ for the nonperiodic part of the orbit, and (iv)–(v) the periodic orbit.

diagram on parameter space (γ, τ) (not shown), similar to the diagram in Figure 11 but with κ not at its homeostasis value.

If the value of κ is changed to $\kappa = 0.662$, but all the other parameters are held at their values from Figure 22, then the chaos becomes persistent. The attractor (not shown) looks very similar to Figure 22(iii), but for $\kappa = 0.662$ the chaos persists through at least 3×10^4 days. That the attractor is of higher dimension than the previous examples can be inferred by comparing how disordered the two-dimensional projection seen in Figure 22(ii) looks compared to the previous examples. The first six Lyapunov exponents are computed numerically to be +0.02444, +0.008055, -0.00004119, -0.006071, -0.01771, and -0.02882. So for this example there are two positive Lyapunov exponents, the sum of the first five exponents is positive, and the Lyapunov dimension of the attractor is d = 5.3. This dimension is relatively high compared to our previous examples and many of the classical examples of chaotic attractors in ODEs, such as the Lorenz attractor [38], for which the dimension is often between 2 and 3. However, DDEs define infinite-dimensional dynamical systems, and it is well-known that they can generate high-dimensional chaotic attractors [46].

5.3. Snaking branch. Continuation in τ with all the other parameters at their values from Table 1 was illustrated in section 3.1 (see Figures 7–8) and appears to show a canard explosion, similar to the canard explosion for γ continuation, described in section 4. Different behavior is observed if we vary all three parameters γ , κ , and τ .

In Figure 23 we present the results of one-parameter continuation in τ with $\gamma = 0.15$, $\kappa = 0.2$, and the other parameters at their values from Table 1. The bifurcation diagram in Figure 23(i) appears to show similar behavior to the earlier τ continuation, with the steady state stable except between a pair of Hopf bifurcations. There is again a subcritical Hopf bifurcation leading to an interval of bistability between the steady state and a stable limit cycle, and there are again ripples in the amplitude and period of solutions along the branch of stable periodic orbits. The period of the orbits but not the amplitude increases significantly to reach 180 days just before the period collapses to 21.2 days at the Hopf bifurcation.

As illustrated in Figure 23(ii) there is *not* a canard this time. Instead the bifurcation branch snakes about 28 times across $\tau = 4.262041$ creating a small interval of τ values for which there are 57 coexisting periodic orbits. If the periodic orbits had been computed just by simulating to only find the stable solutions, it would appear that the amplitude and period both suddenly increase as τ is decreased through 4.262041, suggesting the possibility of a canard explosion. But the DDEBiftool computations, which allow us to compute unstable periodic orbits just as well as stable ones, show this not to be the case.

At the top and bottom of the snake there is a pair of fold bifurcations of periodic orbits both at $\tau \approx 4.262041$, with the τ values of the bifurcations points agreeing to at least seven significant figures. The large amplitude orbit at the top of the snaking branch is stable for very small interval of τ values ($\tau \in (4.262037, 4.262041)$), before losing stability in a period-doubling bifurcation at $\tau \approx 4.262037$. The small amplitude orbit at the bottom of the snake is stable for $\tau \in (4.262037, 4.262054)$, before also losing stability in a period-doubling bifurcation at $\tau \approx 4.262054$. We will come back to the dynamics resulting from these period doublings at the end of this section. On the snaking branch between the two fold bifurcations at $\tau \approx 4.262041$ all the periodic orbits are unstable.



Figure 23. Parameter continuation in τ for periodic orbits with $\gamma = 0.15$, $\kappa = 0.2$. (i) Bifurcation diagram showing Hopf bifurcations at $\tau = 2.21327$ and 4.26817, fold bifurcation of periodic orbits at $\tau = 2.19228$ and 4.262041, and period-doubling bifurcation of periodic orbits at 4.261983, 4.262037, 4.262054, and 4.262183. The inset shows the period of the orbits. (ii) Details of the snaking branch region of the bifurcation diagram from panel (i). (iii) Examples of coexisting periodic orbits for $\tau = 4.26203$ and $\tau = 4.2620395$. (iv) Delay embeddings of coexisting periodic orbits for $\tau = 4.26203$ and inset showing that the orbits are out of phase. (v) Three solutions of (1.1) for $\tau = 4.2620395$ computed using the MATLAB dde23 routine [45] with initial functions given by DDEBiftool solutions for the corresponding coloured dots shown in panel (iii). All three orbits converge to the large amplitude stable limit cycle seen in the bifurcation diagram in panel (iii). (vi) Delay embeddings of the three orbits shown in panel (v) along with the stable limit cycle to which they converge show in blue, also denoted by the blue square in panel (iii).

Figure 23(iv) shows the delay embeddings for the 57 unstable limit cycles that coexist when $\tau = 4.26203$. The positions of these orbits on the snaking branch are indicated in Figure 23(iii), where we use shades of pink to red to indicate orbits which are on the legs of the snake for which the amplitude increases as τ increases and shades of cyan to blue for orbits on the legs of the branch where the amplitude decreases as τ increases. Although Figure 23(iv) is very

reminiscent of Figure 13, there are crucial differences between the dynamics. In particular the orbits shown in Figure 23(iv) are all unstable and all coexist, whereas those of Figure 13 are stable and exist over an exponentially small parameter interval, with a unique orbit existing for each of the parameter value. Nevertheless, there are significant similarities between the dynamics in the two cases with Figure 23(iv) also appearing to indicate the presence of a slow manifold which is stable for a certain range of Q values, with the orbits appearing to spiral onto the slow manifold. The inset in Figure 23(iv) shows that the phase of this convergence is different on the two legs of the snaking branch.

In Figures 23(v)–(vi) we illustrate the dynamics with $\tau = 4.2620395$ when the large amplitude orbit (indicated by the blue dot in Figure 23(iii)) is stable. For three different initial functions corresponding to unstable periodic orbits on the snaking branch (also indicated by colored dots in panel (iii)) we take a part of the periodic orbit generated by DDEBiftool as the initial function, then use the MATLAB dde23 routine [45] to simulate the solution. All three orbits are seen to converge to the stable large amplitude limit cycle, with period about 180 days, but the nature of that convergence is not simple to explain. All of the periodic orbits along with their unstable manifolds are squeezed very close together when the orbits follow the slow manifold before diverging from each other again when the slow manifold becomes unstable, and probably as a consequence of this the connecting orbits between the limit cycles do not appear to have a simple structure. In the figure we see that the orbit shown in orange passes close to the slow manifold many times before approaching the stable limit cycle, around t = 2000 days, while the other two initial functions lead to solutions which converge to the stable periodic orbit relatively quickly.

In Figure 24 we show four orbits located at adjacent local extrema of τ on the snaking branch, as shown in the inset of panel (i). The profiles in panel (i) and delay embeddings in panel (ii) illustrate how the periodic orbit changes along the snaking curve of solutions as the amplitude increases. In Figure 24(i) the phase of the orbits is aligned so that they all have the global maximum and minimum aligned (close to t = 150). Looking back one period to the previous occurrence of the global maxima and minima, we see that the position on the snaking branch of solutions is related to the number of short period oscillations seen as the solution



Figure 24. Example periodic orbits and their delay embedding from the snaking branch shown in Figure 23. The delay embedding shows that oscillations in the $(Q(t), Q(t-\tau))$ projection appear to be in antiphase between the left and right sides of the snaking branch. The orbit profiles show the same behavior when they are plotted with final time points and Q values equal.

converges onto the slow manifold. Crossing each leg of the snaking branch with increasing amplitude corresponds to adding half a short period oscillation to the whole periodic orbit. So the points on the snaking branch at minima of τ display one less/more short oscillation than seen at the next minima of τ directly above/below them on the branch, and half a short period oscillation less/more than seen at the adjacent maxima of τ with larger/smaller amplitude. The time-delay embeddings in Figure 24(ii) and its inset reveal that orbits located in the same extrema of the snaking branch converge to the slow manifold in phase with each other, and in antiphase to orbits located in opposite extrema. Although we have seen how the solution changes along the snaking branch, this does not explain why the branch itself snakes; similar solution behavior but without branch snaking was observed in section 4 for the canard explosion.

As noted near the beginning of the section, at either side of the fold points at $\tau = 4.262041$ there are period-doubling bifurcations. These bifurcations actually come in pairs, resulting in two separate intervals, one each side of the snaking part of the branch, for which the periodic solutions on the principal branch are unstable. For large amplitude solutions this occurs for $\tau \in (4.261983, 4.262037)$ with a period-doubling bifurcation at each end of this interval. For small amplitude solutions, the unstable part of the branch between the period-doubling bifurcations is for $\tau \in (4.262054, 4.262183)$. To explore the dynamics as τ is varied over these parameter intervals in Figure 25 we present an orbit diagram showing the local maxima and minima of Q(t) along the solutions of (1.1) as τ is varied across this region with $\gamma = 0.15$, $\kappa = 0.2$. This is computed similarly to Figure 18, but this time integrating through a transient



Figure 25. Orbit diagram showing local maxima and minima of solutions segments (1.1) as function of the delay τ , with $\gamma = 0.15$, $\kappa = 0.2$ and other parameters taking their values from Table 1. Black and blue dots represent respectively the local maxima and minima for a decreasing sequence of τ values. The bifurcation points of primary branch are identified using the same symbols as in Figures 3 and 15. The inset shows the same interval of τ values as the main figure but for a restricted range of Q values revealing details of the bifurcations.

845

of 1530τ days, then plotting all the maxima and minima that occur over the next 170τ days. A mesh of 2000 equally spaced points for $\tau \in [4.26197, 4.26219]$ was used for decreasing τ . In this case we did not observe any noticeable hysteresis effects. For each mesh the solution over the last τ days was used as the initial history to start the transient computation for the next adjacent τ value.

The results displayed in Figure 25 clearly reveal the bifurcations already shown in Figure 23(ii) including the fold bifurcation near $\tau = 4.262041$ and the period doubling bifurcations at $\tau = 4.261984$, 4.262038, 4.262054, and 4.262184. Between the pairs of period doublings, much richer dynamics are displayed than we had expected. Several period-doubling cascades are clearly visible (in the inset to the figure), leading to several intervals of apparently stable chaotic dynamics. There are also windows of stable periodic dynamics, including a period-3 window, which suggests the possibility of unstable chaotic dynamics (period-3 implies chaos only for one-dimensional maps).

6. Dynamical diseases. In dynamic hematological diseases oscillations are observed in the circulating concentrations of one or more of the cell lines [23]. Mathematical interest has often focused on what have been termed periodic hematological disorders, including cyclic neutropenia (CN), cyclic thrombocytopenia (CT), and periodic chronic myelogenous leukemia (PCML).

CN is one of the most studied of these periodic diseases, with the concentration of circulating neutrophils varying from very low to normal or high levels with a period of about 19 to 21 days [5, 16]. Patients experience a bout of neutropenia (abnormally low neutrophil concentrations) each period, during which time the immune system is impaired and they are more susceptible to infection [20]. For patients with CT, oscillations in platelet counts from normal to very low values are observed with periods between 20 to 40 days [27]. For patients with PCML, cycling in white blood cells from normal to high levels with periods from approximately 30 to 100 days [27] and 40 to 80 days [49] is reported.

Many mathematical models of hematopoiesis have been developed in an effort to understand these diseases and the origins of the oscillatory dynamics [5, 50, 56]. Efforts have often focused on deriving mathematical models and associated parameter sets for which the model has a stable limit cycle with a period commensurate with a particular disease under consideration [15, 16, 37]. Clinical efforts focus in entirely different directions, typically concentrating on alleviating the cytopenia (dangerously low blood cell concentrations) either by raising the concentration nadir or by decreasing the time interval that concentrations are below the recognised cytopenia threshold. Periodic oscillations in the strict mathematical sense are of limited clinical interest, and in the clinical literature the terms periodic and cyclic are often used as synonyms for episodic, and it is not implied that the time intervals between episodes are fixed. Consequently, there are many other hematological disorders which at least for some subjects display dynamics with a periodic signature, but for which there are only scattered case reports of the periodicity. Examples include cyclic 100-day pancytopenia [7], cyclic (approximately 60-day) bicytopenia with Shapiro syndrome [52], and polycythemia vera [48] with approximately 28-day cycling.

CT typically involves oscillations of just the platelets [23], though one case of multilineage CT has recently been reported [36], while for CN and PCML oscillations of all of the major blood cell groups are observed [23]. This suggests that for CN and PCML the cycling in all cell lineages may be due to a dynamic destabilization at the stem cell level [23]. This destabilization occurs through different mechanisms in these two diseases with leukemic HSCs typically presenting a chromosome abnormality in PCML [49]. In CN a mutation in the ELANE gene leads to increased apoptosis in the neutrophil progenitor cells during mitosis [19], and the destabilization of the HSCs appears to be caused by a feedback mechanism from the neutrophil lineage.

Considerable variation in the oscillatory periods is observed within and between these disorders. A Lomb periodogram [37] is typically used to extract a periodic signature from the data, but the data itself is never truly periodic. There can be many reasons for this including data sampling, measurement error, intrinsic stochasticity of cell proliferation and differentiation, environmental variation, adaptation of the model parameters, or simply that the actual dynamics are not periodic.

In Table 1 we gave specific values of the model parameters from which we start our bifurcation analysis. Other authors use somewhat different values, or more correctly report ranges for the parameter values [5, 15]. Through interindividual variability we should expect that a single parameter set will not be suitable for all subjects. However, as seen in section 3 there are no bifurcations near to the stated homeostasis parameters. Hence, using other similar parameter values in the model will also lead to an asymptotically stable steady state.

To provoke a qualitative change in the dynamics of (1.1) requires a large change in the parameters. This situation was already envisioned by Glass and Mackey [25], who coined the term dynamical disease to describe physiological systems where the control system itself is intact, but operating in a parameter range leading to abnormal dynamics. With significant changes to one or more parameters we do observe nontrivial dynamics. These dynamics only become of physiological, rather than mathematical, interest when they produce oscillations with characteristics similar to the reported diseases, and we do observe behavior reminiscent of CN, PCML, and CT.

An increased apoptosis rate γ during the cell cycle, as illustrated in Figures 5 and 6, results in stable oscillations in the HSCs of period between about 75 and 100 days for $\gamma \in [0.2278, 0.24]$. The shortest period orbit illustrated in Figure 6(i) is of interest. This has a maximal value of Q(t) greater than 70% of Q^h , and hence maximum differentiation rate $\kappa Q(t)$ to peripheral blood cell precursors above 70% of the homeostatic rate, while the interval of severely reduced HSC numbers is relatively short (below 4 weeks). Such cycling in the HSCs would naturally result in pancytopenia in a full model of the hematopoietic system. If the apoptosis rate γ is increased slightly above 0.24 longer periodic orbits result, but as seen from Figure 6(ii) these have severely reduced HSC numbers for intervals of hundreds of days, which is much longer than the lifespan of circulating erythrocytes and which would induce a fatal anemia. Still higher values of γ result in complete depletion of the HSCs with Q = 0 becoming the globally attracting stable steady state.

In Figures 3 and 4 we illustrate periodic dynamics of the HSCs for increased values of the differentiation rate κ . The stable periodic orbits in Figure 3 exist when the rate constant κ is 6 or more times its homeostatic value, meaning that in this scenario the rate $\kappa Q(t)$ at which HSCs differentiate to precursors of circulating hematopoietic cells can be elevated compared to the homeostasis value, even when the number of HSCs Q(t) is less than Q^h . While the periods

observed in Figure 3 are too short for PCML, longer periods of 30 to 100 days consistent with PCML can be obtained by also increasing the cell cycle time τ as seen in Figure 15(i) and the inset in Figure 23(i).

The largest periods seen in Figure 3(ii), corresponding to the largest amplitude orbits on the main branch and also the period-doubled orbits have period about 17 days, which is close to but a little shorter than typical periods for CN. Varying three parameters in (1.1) it is possible to find periodic orbits with periods typical of CN, for example, $(\kappa, \gamma, \tau) \approx (0.09, 0.28, 3)$ results in stable limit cycles with period between 19 and 21 days. Stable large amplitude limit cycles are observed in Figure 15 with periods in the 20- to 40-day range typical of CT.

We observed numerous instances of bistability, which allows for the possibility that a therapeutic intervention or some other outside affect on the hematopoietic system could cause it to flip between different stable states. This has been observed in practice, where, for example, G-CSF can induce neutrophil oscillations with a period of about 7 to 15 days for neutropenic individuals [27].

In section 4 we explored a canard explosion. The very long period orbits that we found are likely not physiologically relevant, as they include long time intervals during which the HSCs are severely depleted. During these intervals the production of peripheral blood cells would be so severely compromised that a fatal cytopenia would likely result. Although we do not rule out the possibility that a canard explosion with other parameters might lead to physiologically feasible long period orbits, the singular parameter ε suggests this is unlikely. We see from (4.1) and (4.2) that $\kappa Q^* \sim \varepsilon$, so in the parameter regime $0 < \varepsilon \ll 1$ where we might expect the canard to exist the differentiation of HSCs toward mature blood cell lines will be severely compromised.

The quasi-periodic and chaotic solutions observed in section 5 may be of more physiological relevance for two reasons. First, although these solutions all have significantly reduced HSC concentrations compared to homeostasis, they are found in parameter regions where the differentiation rate κ is significantly increased, so that the differentiation $\kappa Q(t)$ out of the HSC compartment is at or above the homeostatic rate when Q(t) is close to a local maxima. These HSC dynamics would likely lead to episodic pancytopenia in a full model of the hematopoietic system, which could be an interesting topic for follow-up study. A second reason why these dynamics are of physiological relevance is that they show the system generating nonconstant nonperiodic dynamics which is more akin to what is seen in real data than the purely periodic solutions that we investigated earlier.

In this section we highlighted some of the solutions that we observed with periods in ranges characteristic of dynamical diseases. The two-parameter continuations of section 3.2 could be used as a starting point for an extended study to find additional parameter regions with periodic solutions commensurate with dynamical diseases. Although it would be tractable to do that for the HSC model (1.1), such a study would be more interesting in a model of the hematopoietic system that incorporates multiple mature cell lines. We have clearly shown that our HSC model can demonstrate the oscillatory dynamics characteristic of dynamical diseases, without the need for any feedback loops from more mature cell lines. However, many of the solutions with interesting dynamics are associated with an increased differentiation rate κ . It remains an open question in particular dynamical diseases whether the differentiation rate is actually raised, and if so whether this is intrinsic to the disease-state HSC dynamics or caused by feedback from the peripheral blood cell dynamics.

7. Discussion and conclusions. We set out to show that the HSC model (1.1) could generate limit cycles of periods typical in dynamical diseases, simply by changing some of the parameter values in the model. Long period orbits had previously been observed by varying s [40, 49, 50]. We varied the parameters γ , κ , and τ and found periodic orbits of periods from about 1 week up to 9 years, encompassing the 19 to 21 days typical of CN, 20 to 40 days of CT, and the 30 to 100 days of PCML. Whereas the model (1.1) treats the HSCs as a single homogeneous population, more recent mathematical models couple multiple copies of (1.1) together [2, 51] to represent the different maturity levels of HSCs and should be able to generate similar dynamics.

We also observed a plethora of more exotic dynamics including mixed mode oscillations, period-doubling cascades, and chaotic solutions. In section 5.1 we showed that the DDE (1.1) admits stable torus solutions. Elsewhere, in section 4 we studied a putative canard explosion, identified the singular variable, and constructed an approximation to the slow manifold and nearby dynamics. We showed that the local stability of the slow manifold changes very close to the point where the stability of the critical manifold changes. Our analysis of the canard explosion is incomplete. Established analysis and constructions rely on separating the slow and fast variables [55]. In contrast, (1.1) is scalar and does not have a simple natural separation into fast and slow subsystems. We believe this to be the first demonstration of canard-like behavior in a scalar system, and a full analysis will require an extension to current theory. In the current work we present a detailed numerical investigation of the phenomenon, with the hope that it will intrigue the theoreticians to complete the analysis.

Equation (1.1) clearly displays mixed mode oscillations (see the time plots in Figures 6, 8, and 12). Such dynamics are usually associated with slow-fast systems and coupled oscillators, and it is rather curious to see these phenomena in the scalar DDE (1.1). It is well-known that such dynamics are possible when there are multiple delays, and in the case of two state-dependent delays no other nonlinearity is required other than the state-dependency of the delays [12]. In that case it seems that essentially the two delay terms interact as if they are coupled oscillators. However, (1.1) is scalar with only one delay and has no Hopf-Hopf bifurcations. Equation (1.1) is in the general class of problems

$$\dot{u}(t) = -u(t) - \alpha h(u(t)) + Ah(u(t-\tau)),$$

where h(u) is a unimodal function. Problems of this form have been studied in the case $\alpha = 0$, but we are not aware of systemic theoretical studies of the more general case with $\alpha \neq 0$. It seems likely to us that the dynamics reminiscent of coupled oscillators are generated by an interaction between the two instances of the nonlinearity evaluated at the current time t and the delayed time $t - \tau$.

We found many examples of bistability in (1.1). These include bistability between pairs of periodic orbits (Figures 3 and 15), a periodic orbit and a stable steady state (Figures 3, 5, 7, and 15), a periodic orbit and a torus (Figures 15 and 16), as well as bistability between chaotic and nonchaotic solutions (Figure 18). Bistability of periodic orbits is caused by pairs of fold bifurcations of limit cycle which originate in cusp bifurcations (seen in Figures 9 and 10).

DYNAMICS OF A HEMATOPOIETIC STEM-CELL MODEL

We found both subcritical and supercritical Hopf bifurcations and bistability between a stable limit cycle and a steady state is associated with the Bautin or generalized Hopf bifurcation (denoted in Figures 9 and 11) where the criticality of the Hopf bifurcation changes. A curve of fold bifurcations of limit cycles emerges from this point, which results in the interval of bistability seen in the one-parameter continuations between the fold and the subcritical Hopf bifurcation. Few previous studies have been sufficiently detailed to detect the criticality of the Hopf bifurcations, but those that were only found supercritical Hopf bifurcations [5, 6, 47], though Bernard, Bélair, and Mackey [5] did find Hopf bifurcations which were close to a criticality change.

Mathematical studies of differential equations that model hematopoiesis frequently focus on existence and stability of a nontrivial solution. Once a Hopf bifurcation is found, the steady state becomes unstable, and secondary bifurcations to more complex dynamical structures are often not pursued. On the other hand, peripheral blood samples are often only taken for a few days at a time during a hospital stay and are otherwise not taken, or are taken at widely and irregular spaced intervals. The data, even when well sampled, frequently appears noisy, and it is unheard of to see solutions that are exactly periodic. Often, a periodic signature is only revealed by a frequency test, such as the Lomb periodogram. In this context, the bistable, long period, quasi-periodic, and transient and persistent chaotic dynamics that we find are very interesting. The same individual can present very different looking dynamics during different sampling intervals. It might be that the dynamics are actually periodic, but the period is so long that different parts of the periodic solution are revealed by different sampling intervals. Another possibility is that the dynamics are actually chaotic (but not random), and different parts of the chaotic attractor are revealed at different times. As we saw in Figure 20 a time series of chaotic dynamics can appear to be surprisingly regular, while in Figure 18 where the time series of the dynamics was clearly not regular, the system actually spends significant time near the attractor of the first example. In a period-doubling cascade to chaos the initial seed orbit and its low order period doublings continue to exist after they become unstable and can be expected to have some organizing influence on the structure of the dynamics. Thus it is natural to expect there to be some periodic signal contained in the time series, even of a chaotic solution, and it is very unlikely that sufficient blood measurements would be taken from a single subject to discern genuinely chaotic dynamics. A widely used strategy for determining perturbed parameters associated with dynamical diseases is to try to find parameters which generate a periodic solution which is closest to the data [37]. Given the difficulty in distinguishing between chaotic and periodic dynamics, and the ability of the mathematical models to generate both, this strategy may not be optimal, and we should consider that the disease dynamics may not generate a simple periodic orbit but that there may be multiple bifurcations between the homeostasis and diseased states, leading to more complex dynamics.

Acknowledgments. We are grateful to Mike Mackey for useful discussions and feedback on a draft of this work. We also thank John Mitry for our discussions on canards. We are grateful to Dimitri Breda for sharing his code for the computation of Lyapunov exponents in DDEs. Finally, we wish to thank the referees and editor for their many constructive suggestions.

REFERENCES

- M. ADIMY, F. CRAUSTE, AND A. EL ABDLLAOUI, Asymptotic behavior of a discrete maturity structured system of hematopoietic stem cell dynamics with several delays, Math. Model. Nat. Phenom., 1 (2006), pp. 1–22, https://doi.org/10.1051/mmnp:2008001.
- M. ADIMY, F. CRAUSTE, AND S. RUAN, Periodic oscillations in leukopoiesis models with two delays, J. Theoret. Biol., 242 (2006), pp. 288–299, https://doi.org/10.1016/j.jtbi.2006.02.020.
- [3] L. K. ANDERSEN AND M. C. MACKEY, Resonance in periodic chemotherapy: A case study of acute myelogenous leukemia, J. Theoret. Biol., 209 (2001), pp. 113–130, https://doi.org/10.1006/jtbi.2000. 2255.
- [4] E. BENOÎT, J. L. CALLOT, F. DIENER, AND M. DIENER, Chasse au canard, Collect. Math., 32 (1981), pp. 37–119.
- [5] S. BERNARD, J. BÉLAIR, AND M. C. MACKEY, Oscillations in cyclical neutropenia: New evidence based on mathematical modeling, J. Theoret. Biol., 223 (2003), pp. 283–298, https://doi.org/10.1016/ S0022-5193(03)00090-0.
- S. BERNARD, J. BÉLAIR, AND M. C. MACKEY, Bifurcations in a white-blood-cell production model, C. R. Biol., 327 (2004), pp. 201–210, https://doi.org/10.1016/j.crvi.2003.05.005.
- H. S. BIRGENS AND H. KARLE, Reversible adult-onset cyclic haematopoiesis with a cycle length of 100 days, Br. J. Haematol., 83 (1993), pp. 181–186, https://doi.org/10.1111/j.1365-2141.1993.tb08269.x.
- [8] D. BREDA, S. MASET, AND R. VERMIGLIO, Stability of Linear Delay Differential Equations. A Numerical Approach with MATLAB, Springer, New York, 2015.
- D. BREDA AND E. S. VAN VLECK, Approximating Lyapunov exponents and Sacker-Sell spectrum for retarded functional differential equations, Numer. Math., 126 (2014), pp. 225–257, https://doi.org/ 10.1007/s00211-013-0565-1.
- [10] H. W. BROER, G. B. HUITEMA, AND M. B. SEVRYUK, Quasi-Periodic Motions in Families of Dynamical Systems. Order Amidst Chaos, Lecture Notes in Math. 1645, Springer, New York, 1996.
- [11] F. J. BURNS AND J. F. TANNOCK, On the existence of a G₀-phase in the cell cycle, Cell Tissue Kinet., 3 (1970), pp. 321–334, https://doi.org/10.1111/j.1365-2184.1970.tb00340.x.
- [12] R. C. CALLEJA, A. R. HUMPHRIES, AND B. KRAUSKOPF, Resonance phenomena in a scalar delay differential equation with two state-dependent delays, SIAM J. Appl. Dyn. Syst., 16 (2017), pp. 1474–1513, https://doi.org/10.1137/16M1087655.
- [13] S. A. CAMPBELL, E. STONE, AND T. ERNEUX, Delay induced canards in a model of high speed machining, Dyn. Syst., 24 (2009), pp. 373–392, https://doi.org/10.1080/14689360902852547.
- [14] C. COLIJN, A. C. FOWLER, AND M. C. MACKEY, High frequency spikes in long period blood cell oscillations, J. Math. Biol., 53 (2006), pp. 499–519, https://doi.org/10.1007/s00285-006-0027-9.
- [15] C. COLIJN AND M. C. MACKEY, A mathematical model of hematopoiesis: I. Periodic chronic mylogenous leukemia, J. Theoret. Biol., 237 (2005), pp. 117–132, https://doi.org/10.1016/j.jtbi.2005.03.033.
- [16] C. COLIJN AND M. C. MACKEY, A mathematical model of hematopoiesis: II. Cyclical neutropenia, J. Theoret. Biol., 237 (2005), pp. 133–146, https://doi.org/10.1016/j.jtbi.2005.03.034.
- [17] R. M. CORLESS, G. H. GONNET, D. E. G. HARE, D. J. JEFFREY, AND D. E. KNUTH, On the Lambert W function, Adv. Comput. Math., 5 (1996), pp. 329–359, https://doi.org/10.1007/BF02124750.
- [18] M. CRAIG, A. R. HUMPHRIES, AND M. C. MACKEY, A mathematical model of granulopoiesis incorporating the negative feedback dynamics and kinetics of G-CSF/neutrophil binding and internalization, Bull. Math. Biol., 78 (2016), pp. 2304–2357, https://doi.org/10.1007/s11538-016-0179-8.
- [19] D. C. DALE, A. A. BOLYARD, AND A. APRIKYAN, Cyclic neutropenia, Semin. Hematol., 39 (2002), pp. 89–94, https://doi.org/10.1053/shem.2002.31917.
- [20] D. C. DALE AND K. WELTE, Neutropenia and neutrophilia, in Williams Hematology, 9th ed., K. Kaushansky et al. eds., McGraw-Hill, New York, 2016.
- [21] K. ENGELBORGHS, T. LUZYANINA, AND D. ROOSE, Numerical bifurcation analysis of delay differential equations using DDE-BIFTOOL, ACM Trans. Math. Software, 28 (2002), pp. 1–21, https://doi.org/ 10.1145/513001.513002.
- [22] N. FENICHEL, Geometric singular perturbation theory for ordinary differential equations, J. Differential Equations, 31 (1979), pp. 53–98, https://doi.org/10.1016/0022-0396(79)90152-9.

- [23] C. FOLEY AND M. C. MACKEY, Dynamic hematological disease: A review, J. Math. Biol., 58 (2009), pp. 285–322, https://doi.org/10.1007/s00285-008-0165-3.
- [24] A. C. FOWLER AND M. C. MACKEY, Relaxation oscillations in a class of delay differential equations, SIAM J. Appl. Math., 63 (2002), pp. 299–323, https://doi.org/10.1137/S0036139901393512.
- [25] L. GLASS AND M. C. MACKEY, Pathological conditions resulting from instabilities in physiological control systems, Ann. N. Y. Acad. Sci., 316 (1979), pp. 214–235, https://doi.org/10.1111/j.1749-6632.1979. tb29471.x.
- [26] J. K. HALE AND S. M. VERDUYN LUNEL, Introduction to Functional Differential Equations, Appl. Math. Sci. 99, Springer, New York, 1993.
- [27] C. HAURIE, D. C. DALE, AND M. C. MACKEY, Cyclical neutropenia and other periodic hematological disorders: A review of mechanisms and mathematical models, Blood, 92 (1998), pp. 2629–2640.
- [28] N. D. HAYES, Roots of the transcendental equation associated with a certain difference-differential equation, J. London Math. Soc., s1-25 (1950), pp. 226–232, https://doi.org/10.1112/jlms/s1-25.3.226.
- [29] T. INSPERGER AND G. STÉPÁN, Semi-Discretization for Time-Delay Systems, Springer, New York, 2011.
- [30] E. IZHIKEVICH, Dynamical Systems in Neuroscience: The Geometry of Excitability and Bursting, MIT Press, Cambridge, MA, 2007.
- [31] J. L. KAPLAN AND J. A. YORKE, Chaotic behavior of multidimensional difference equations, in Functional Differential Equations and Approximation of Fixed Points: Proceedings, Bonn, July 1978, H.-O. Peitgen and H.-O. Walther, eds., Springer, New York, 1979, pp. 204–227, https://doi.org/10.1007/ BFb0064319.
- [32] K. KAUSHANSKY, Hematopoietic stem cells, progenitors, and cytokines, in Williams Hematology, 9th ed., K. Kaushansky et al., eds., McGraw-Hill, New York, 2016.
- [33] V. KOLMANOVSKII AND A. MYSHKIS, Introduction to the Theory and Applications of Functional Differential Equations, Math. Appl., Kluwer Academic Publishers, Dordrecht, 1999.
- [34] M. KRUPA AND J. D. TOUBOUL, Canard explosion in delay differential equations, J. Dynam. Differential Equations 28 (2016), pp. 471–491, https://doi.org/10.1007/s10884-015-9478-2.
- [35] Y. A. KUZNETSOV, Elements of Applied Bifurcation Theory, 3rd ed., Appl. Math. Sci. 112, Springer, New York, 2004.
- [36] G. P. LANGLOIS, D. M. ARNOLD, J. POTTS, B. LEBER, D. C. DALE, AND M. C. MACKEY, Cyclic thrombocytopenia with statistically significant neutrophil oscillations, Clin. Case. Rep., 6 (2018), pp. 1347–1352, https://doi.org/10.1002/ccr3.1611.
- [37] G. P. LANGLOIS, M. CRAIG, A. R. HUMPHRIES, M. C. MACKEY, J. M. MAHAFFY, J. BÉLAIR, T. MOULIN, S. R. SINCLAIR, AND L. WANG, Normal and pathological dynamics of platelets in humans, J. Math. Biol., 75 (2017), pp. 1411–1462, https://doi.org/10.1007/s00285-017-1125-6.
- [38] E. N. LORENZ, Deterministic nonperiodic flow, J. Atmospheric Sci., 20 (1963), pp. 130–141.
- [39] M. C. MACKEY, Unified hypothesis for the origin of aplastic anemia and periodic haematopoiesis, Blood, 51 (1978), pp. 941–956.
- [40] M. C. MACKEY, C. OU, L. PUJO-MENJOUET, AND J. WU, Periodic oscillations of blood cell populations in chronic myelogenous leukemia, SIAM J. Math. Anal., 38 (2006), pp. 166–187, https://doi.org/10. 1137/04061578X.
- [41] M. C. MACKEY AND R. RUDNICKI, Global stability in a delayed partial differential equation describing cellular replication, J. Math. Biol., 33 (1994), pp. 89–109, https://doi.org/10.1007/BF00160175.
- [42] J. MALLET-PARET AND R. D. NUSSBAUM, Global continuation and asymptotic behavior for periodic solutions of a differential-delay equation, Ann. Mat. Pura Appl., 145 (1986), pp. 33–128, https://doi. org/10.1007/BF01790539.
- [43] J. MALLET-PARET AND R. D. NUSSBAUM, Boundary layer phenomena for differential-delay equations with state-dependent time lags: III, Discrete Contin. Dyn. Syst. Ser. A, 189 (2003), pp. 640–692, https://doi.org/10.1016/S0022-0396(02)00088-8.
- [44] J. MALLET-PARET AND R. D. NUSSBAUM, Superstability and rigorous asymptotics in singularly perturbed state-dependent delay-differential equations, J. Differential Equations, 250 (2011), pp. 4037–4084, https://doi.org/10.1016/j.jde.2010.10.024.
- [45] MATHWORKS, MATLAB 2015b, Mathworks, Natick, MA, 2015.
- [46] B. MENSOUR AND A. LONGTIN, Power spectra and dynamical invariants for delay-differential and difference equations, Phys. D, 113 (1998), pp. 1–25, https://doi.org/10.1016/S0167-2789(97)00185-1.

- [47] J. G. MILTON AND M. C. MACKEY, Periodic haematological diseases: Mystical entities or dynamical disorders?, J. Roy. Coll. Phys. (Lond.), 23 (1989), pp. 236–241.
- [48] A. MORLEY, Blood-cell cycles in polycythæmia vera, Australasian Ann. Medicine, 18 (1969), pp. 124–126, https://doi.org/10.1111/imj.1969.18.2.124.
- [49] L. PUJO-MENJOUET, S. BERNARD, AND M. C. MACKEY, Long period oscillations in a G₀ model of hematopoietic stem cells, SIAM J. Appl. Dyn. Syst., 4 (2005), pp. 312–332, https://doi.org/10.1137/ 030600473.
- [50] L. PUJO-MENJOUET AND M. C. MACKEY, Contribution to the study of periodic chronic myelogenous leukemia, C. R. Biol., 327 (2004), pp. 235–244, https://doi.org/10.1016/j.crvi.2003.05.004.
- [51] Y. QU, J. WEI, AND S. RUAN, Stability and bifurcation analysis in hematopoietic stem cell dynamics with multiple delays, Phys. D, 239 (2010), pp. 2011–2024, https://doi.org/10.1016/j.physd.2010.07.013.
- [52] L. E. ROEKER, V. GUPTA, W. I. GONSALVES, A. P. WOLANSKYJ, AND N. GANGAT, Cyclic bicytopenia in a patient with shapiro syndrome, Case Rep. Hematol., 2013 (2013), 231713, https://doi.org/10. 1155/2013/231713.
- [53] J. SIEBER, K. ENGELBORGHS, T. LUZYANINA, G. SAMAEY, AND D. ROOSE, DDE-BIFTOOL Manual-Bifurcation Analysis of Delay Differential Equations, arXiv:1406.7144 [math.DS], 2015.
- [54] H. SMITH, An Introduction to Delay Differential Equations with Applications to the Life Sciences, Springer, New York, 2010.
- [55] M. WECHSELBERGER, J. MITRY, AND J. RINZEL, *Canard theory and excitability*, in Nonautonomous Dynamical Systems in the Life Sciences, P. Kloeden and C. Pötzsche, eds., Springer, New York, 2013.
- [56] C. ZHUGE, M. C. MACKEY, AND J. LEI, Origins of oscillation patterns in cyclical thrombocytopenia, J. Theoret. Biol., 462 (2019), pp. 432–445, https://doi.org/10.1016/j.jtbi.2018.11.024.