# One-dimensional unstable eigenfunction and manifold computations in delay differential equations 

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#### Abstract

In this paper we present a new numerical technique for computing the unstable eigenfunctions of a saddle periodic orbit in a delay differential equation. This is used to obtain the necessary starting data for an established algorithm for computing one-dimensional (1D) unstable manifolds of an associated saddle fixed point of a suitable Poincaré map. To illustrate our method, we investigate an intermittent transition to chaos in a delay system describing a semiconductor laser subject to phase-conjugate feedback.


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## 1. Introduction

Numerical tools for the study of delay differential equations (DDEs) are of much interest to applied fields because models featuring a delay term appear in applications ranging from chemistry [8] and biology $[1,26]$ to laser physics [21,28]. Until quite recently, the only techniques available to study DDEs were simulation by direct numerical integration of the DDE, or a linear stability analysis of steady states. This is now changing with the introduction of new tools for the numerical bifurcation analysis of DDEs. At the fore of this new software is the continuation package DDE-BIFTOOL [7] allowing the user to find and follow steady states and periodic solutions in systems of DDEs irrespective of their stability; first examples of continuation studies with DDE-BIFTOOL can be found in [5,13-17,27].

[^0]In [20] we introduced the first method for computing one-dimensional (1D) unstable manifolds of a saddle periodic orbit of a DDE with one unstable Floquet multiplier. Specifically, we compute each of the two branches of the unstable manifold $W^{u}(q)$ of a saddle fixed point $q$ of a suitable Poincaré map. This method requires as starting data the saddle point $q$ and two first points $q_{\delta}^{ \pm}$(one on each branch) approximately on $W^{u}(q)$ close to $q$. Using linear approximation near $q$, the points $q_{\delta}^{ \pm}$can be chosen some distance $\delta$ away from $q$ along the unstable eigenspace $E^{u}(q)$. The saddle periodic orbit $q$ can readily be found with DDE-BIFTOOL. However, for a DDE finding a good approximation of the first points $q_{\delta}^{ \pm}$is more difficult. This is the subject of this paper.

In [20] we used an iterative power method to find a vector $v$ approximately spanning the linear unstable eigenspace $E^{u}(q)$. This approach was also used in the detailed study of [15] for a DDE describing a semiconductor laser subject to phase-conjugate feedback (PCF), the example also used here and fully introduced in Section 4. The power method works in the examples in [15,20]. By computing the invariant torus on which the dynamics of the PCF laser are locked, we were able to show that there is a transition to chaos culminating in a crisis bifurcation [15]. However, this iterative power method has its shortcomings; see Section 3 for details. In particular, we found that we were unable to compute both branches of the unstable manifold near bifurcations of the saddle periodic orbit, and in particular, close to some regions of chaotic dynamics.

This realisation motivated the present paper. We present here a method for obtaining the entire starting data for a manifold computation directly from a DDE-BIFTOOL computation. Specifically, we modified DDE-BIFTOOL to produce the unstable eigenfunction associated with a single unstable Floquet multiplier, this data is then used to find the points $q_{\delta}^{ \pm}$. This is less straightforward than it sounds owing to the infinite phase space of the DDE, and it requires several steps to get from the 'raw data' of DDE-BIFTOOL to a suitable representation of $q_{\delta}^{ \pm}$.

To illustrate our technique we consider a sudden transition to chaos in the PCF laser caused by a saddlenode bifurcation of periodic orbits. In [19] this transition was found and identified by simulation and in [15] continuation techniques were used to follow periodic orbits to the saddle-node bifurcation. It appeared that the sudden transition to chaos is an intermittent transition [11,25], characterised by the saddle-node bifurcation taking place 'on a chaotic attractor'. By this we mean that one branch of the 1 D unstable manifold of the saddle resembles the chaotic attractor that existed before the saddle-node bifurcation. Here we confirm the presence of an intermittent transition in a DDE for the first time by computing the unstable manifold of the saddle periodic orbit when this bifurcation is approached. This is only possible because we can now find much better starting data for the 1D manifold algorithm.

The paper is organised as follows. In Section 2 we introduce some background on the theory of DDEs. In Section 3 we explain how we find the starting data. Specifically, in Section 3.1 we detail how DDEBIFTOOL is used to compute unstable eigenfunctions and in Section 3.2 we show how this eigenfunction data is manipulated to obtain the starting data for a manifold computation. As an illustration we show in Section 4 an intermittent transition to chaos in the PCF laser. Finally, we draw conclusions and discuss future work in Section 5.

## 2. Delay differential equations

We now briefly recall some basic facts on the theory of DDEs; see [3,18,29] for further details. Readers may find it useful to look ahead to the concrete example of the PCF laser, system (12) in Section 4. We consider the simplest case, namely an autonomous DDE with a single fixed delay. It has the general form

$$
\begin{equation*}
\frac{\mathrm{d} x(t)}{\mathrm{d} t}=F(x(t), x(t-\tau), \lambda), \tag{1}
\end{equation*}
$$

where $F: \mathbb{R}^{n} \times \mathbb{R}^{n} \times \mathbb{R}^{p} \rightarrow \mathbb{R}^{n}$ is differentiable and $\tau \in \mathbb{R}^{>0}$ is a fixed delay, while $\lambda \in \mathbb{R}^{p}$ is a multiparameter.

We call $\mathbb{R}^{n}$ the physical space of the system. The phase space of (1) is the infinite-dimensional space of continuous functions $\mathscr{C}$ defined on the interval $[-\tau, 0]$ with values in $\mathbb{R}^{n}$. A point, say $q \in \mathscr{C}$, lives in this infinite-dimensional phase space, that is, $q$ is a continuous function

$$
\begin{equation*}
q:[-\tau, 0] \rightarrow \mathbb{R}^{n} . \tag{2}
\end{equation*}
$$

We call $q(0)$ the headpoint of $q$ and $\left.q\right|_{[-\tau, 0)}=\{q(t) \mid t \in[-\tau, 0)\}$ its history.
The evolution of a point $q \in \mathscr{C}$ after time $t \geqslant 0$ is given by the evolution operator

$$
\begin{equation*}
\Phi^{t}: \mathscr{C} \rightarrow \mathscr{C} . \tag{3}
\end{equation*}
$$

This is formally given by an abstract differential equation on the infinite-dimensional phase space [4].
A periodic solution $\Gamma(t)$ of (1) is a solution that repeats itself after some period $\mathrm{T}>0$, that is, $\Gamma(t)=\Gamma(t+\mathrm{T})$ for all $t$. Each segment $q \in \mathscr{C}$ of $\Gamma$, that is, $q(\theta)=\Gamma(t+\theta)$ for $\theta \in[-\tau, 0]$, is a periodic point of the evolution operator $\Phi^{\top}$, such that $\Phi^{\top}(q)=q$. The periodic orbit $\Gamma$ traces out a closed curve in projection onto the physical space $\mathbb{R}^{n}$; see already Fig. 2 for a concrete example.

For a prescribed section $\Sigma \subset \mathbb{R}^{n}$ and by denoting $\mathscr{C}_{\Sigma}$ as the space of points in $\mathscr{C}$ with headpoints in $\Sigma$, the Poincaré map $P$ is defined as

(a)

(b)

(c)

Fig. 1. Time traces over one period T of the components $E_{x}(\mathrm{a}), E_{y}(\mathrm{~b})$ and $N(\mathrm{c})$ of the saddle periodic orbit $\Gamma$ together with $\Gamma+\delta y_{\mu}$ for $\delta=3.0$, illustrating one branch of the unstable linear eigenspace $E^{u}(\Gamma)$.

$$
\begin{equation*}
P: \mathscr{C}_{\Sigma} \rightarrow \mathscr{C}_{\Sigma}, \quad q \mapsto \Phi^{t_{q}}(q) \tag{4}
\end{equation*}
$$

where $t_{q}>0$ is the return time to $\Sigma$. After choosing a section $\Sigma$ (locally) transverse to a periodic orbit $\Gamma$, the point $q \in \mathscr{C}_{\Sigma}$ is a fixed point under $P$ (and a periodic point of (1)).

Near $q$ the map $P$ can be defined as the $j$ th return to $\Sigma$ for some fixed $j$, where $j$ counts all intersections of $\Gamma$ with $\Sigma$ (of which $j-1$ are outside a small neighbourhood of $q$ ). We note that it is generally not possible to define $P$ globally as the $j$ th return map to $\Sigma$ for a fixed $j$, that is, the flow may fail to be transverse and this changes the number of returns to $\Sigma$. (This is in contrast to periodically forced systems, which do have a globally defined Poincaré map in the form of the stroboscopic map of the forcing frequency [2].) Such a tangency can occur at the start point, an interior point or the end point of the orbit of the flow. We will encounter such tangencies in Section 4.

The stability of $\Gamma$ is given by its Floquet multipliers, which are the solutions of a transcendental eigenvalue problem given by the linearisation of (1) around $\Gamma$; see Section 3.1. This linearisation is directly connected to the linearisation $D P(q)$ of the Poincaré map $P$ at the respective fixed point $q \in \mathscr{C}_{\Sigma}$ : the Floquet multipliers are the eigenvalues of $D P(q)$. We remark that the only exception to this is the so-called 'trivial' Floquet multiplier at +1 of $\Gamma$, which is always present and corresponds to the tangent direction to $\Gamma$, but does not appear as an eigenvalue of $D P(q)$. It is a crucial property of DDEs (with fixed delay) that $D P(q)$ is a compact operator,


Fig. 2. The saddle periodic orbit $\Gamma$ from Fig. 1 together with $\Gamma+\delta y_{\mu}$ in panels (a) and (b), and with $\Gamma-\delta y_{\mu}$ in panels (c) and (d), all for $\delta=3.0$; shown in projection onto $(E, N)$-space and $E$-space, respectively. Also shown are the points $q_{\delta}^{ \pm}$for $\delta=3.0$.
which implies that its spectrum consists of countably many eigenvalues (the Floquet multipliers) with the origin of the complex plane as their only possible accumulation point; see [18,29] for more details. In other words, for any fixed $r>0$ there are only a finite number of Floquet multipliers outside a circle of radius $r$, so that there are always only a finite number of unstable eigendirections (associated with Floquet multipliers outside the unit circle). A periodic orbit is called hyperbolic if there are no Floquet multipliers on the unit circle (except for the trivial multiplier). A hyperbolic periodic orbit is either stable if all the Floquet multipliers are inside the unit circle or of saddle type with finitely many unstable eigendirections.

If there are other Floquet multipliers (in addition to the trivial multiplier at +1 ) on the unit circle then the system is undergoing a bifurcation. In this study we will encounter a saddle-node bifurcation of limit cycles which is associated with a real Floquet multiplier crossing the unit circle at +1 .

In what follows, we are interested in the one-dimensional unstable manifold $W^{u}(q)$ of a saddle point $q \in \mathscr{C}_{\Sigma}$ associated with a periodic orbit $\Gamma$, with exactly one unstable Floquet multiplier. $W^{u}(q)$ is the set of all points $p \in \mathscr{C}_{\Sigma}$ that can be iterated backwards under $P$ and are such that $P^{l}(p) \rightarrow q$ as $l \rightarrow-\infty$. At $q$ the unstable manifold $W^{u}(q)$ is tangent to the linear eigenspace $E^{u}(q)$ spanned by the unstable eigenfunction. In projection onto the physical space $\mathbb{R}^{n}$, the 1D linear eigenspace $E^{u}(q)$ forms a one-parameter family of directions along (the history of) $q$; see already Figs. 1 and 2.

In projection onto the physical space $\mathbb{R}^{n}$ a 1 D unstable manifold $W^{u}(q)$ forms a complicated object. However, its trace $W^{u}(q) \cap \Sigma$ is a 1D curve that is smooth (except possibly at isolated points due to the projection). It can be interpreted in much the same way as a 1 D unstable manifold of a fixed point of a planar map. We remark that, as an artifact of the projection, the trace may have self-intersections (reminding one of the fact that $W^{u}(q)$ lives in an infinite-dimensional phase space).

## 3. Unstable eigenfunctions and manifold computations

The method presented in [20] for computing 1D unstable manifolds of saddle periodic orbits in DDEs was developed from that in [22] for maps. It grows the manifold as a sequence of points $\left\{p_{k}\right\}$, which are all in $\mathscr{C}_{\Sigma}$, where we use linear approximation between neighbouring points. The distance between these points is governed by the curvature of the trace of the manifold, which is given by the sequence $\left\{p_{k}(0)\right\}$ of headpoints. (Because the distance between $p_{k}(0)$ and $p_{k+1}(0)$ goes to zero with the prescribed accuracy parameters and the local interpolation error between $p_{k}(t)$ and $p_{k+1}(t)$ depends continuously on $t$, the overall interpolation error is bounded; see [20] for details. Alternatively, one could check the curvature pointwise between three consecutive points for all points along the history array and consider the maximum. However, this is computationally more expensive and presently not implemented.)

Supposing that the manifold has been computed up to the point $p_{k}$, the idea is to find the next point $p_{k+1}$ such that the headpoint $p_{k+1}(0)$ is a distance $\Delta_{k}$ from the headpoint $p_{k}(0)$. This is done by finding a preimage $\hat{p}$ of the point $p_{k+1}$ which lies on the computed part of the manifold. By identifying the two points $p_{l}$ and $p_{l+1}$ between which $\hat{p}$ must lie, $\hat{p}$ is found by bisection. To reduce the number of bisection steps, a small tolerance $\varepsilon$ is allowed, meaning that we only require that

$$
\begin{equation*}
(1-\varepsilon) \Delta_{k}<\left|P(\hat{p}(0))-p_{k}(0)\right|<(1+\varepsilon) \Delta_{k} . \tag{5}
\end{equation*}
$$

The distance $\Delta_{k}$ is adapted during computations according to the curvature of the trace as pre-specified by four accuracy parameters; introduced fully in [20]. A computation stops after a prescribed arclength of the trace has been reached, or when $\Delta_{k}$ falls below a pre-specified small value, thus detecting convergence to an attracting fixed point.

The starting data needed for a manifold computation is a saddle fixed point $q \in \mathscr{C}_{\Sigma}$ of the Poincaré map and two points along $W^{u}(q)$, one on each of the two branches of $W^{u}(q)$. We work here with the linear approximation to $W^{u}(q)$ near $q$ and need to find points $q_{\delta}^{ \pm} \in \mathscr{C}_{\Sigma}$ at a distance $\delta$ from $q$ along the linear
unstable eigenspace $E^{u}(q)$. In particular, this means that $q_{\delta}^{ \pm}(0) \in \Sigma$. The question is: how can one find $q_{\delta}^{ \pm}$ numerically?

Naively, any initial condition near to the point $q$ could be used as starting data to compute one branch of the unstable manifold. This is due to the 'forgiving nature' of the problem, that is, for a saddle fixed point $q$ with one unstable Floquet multiplier the unstable manifold is locally attracting. However, in general we need to find both branches of the unstable manifold which lie on either side of the infinite-dimensional stable manifold, an object which we cannot compute. In [20], motivated by the ordinary differential equation case, we used an iterative power method to find a vector $v \in \mathscr{C}_{\Sigma}$ approximately spanning $E^{u}(q)$. Specifically, in the power method we choose a starting point close to $q$ and perform a small number of iterations of the Poincaré map. The resulting point $v$ is then normalised so that $|v(t)|=1$ for all $t \in[-\tau, 0]$. As starting data we used $q$ and $q \pm \delta v$. This works quite well and one obtains starting data for both branches of the unstable manifold $W^{u}(q)$ when the system is not too close to a bifurcation of the saddle point $q$, as was the case for the examples in $[15,20]$.

However, the vector $v$ does not span $E^{u}(q)$. First of all, the magnitude of the eigenfunction spanning $E^{u}(q)$ varies along $\Gamma$, while for the iterative approximation $v$ it is constant; see already Fig. 1. Second, $v(0)$ lies in $\Sigma$, but in general, $E^{u}(q)$ has a component pointing out of the section $\Sigma$; see already Fig. 2. As a result, the typical problem arises that both $q(t)+\delta v(t)$ and $q(t)-\delta v(t)$ lie on the same side of the stable manifold even for very small $\delta$, so that only one branch of $W^{u}(q)$ can be computed.

The shortcomings of the vector $v$ obtained by the power method motivated the work in this paper. We now explain how reliable starting data for a 1D manifold computation can be found in the form of the two first points $q_{\delta}^{ \pm}$along $E^{u}(q)$ as defined above. This guarantees that we find both branches of the unstable manifold. To this end, we directly compute the eigenvectors of the characteristic matrix with the package DDE-BIFTOOL. As this matrix is already available in DDE-BIFTOOL, the extra computational cost is negligible. Furthermore, using DDE-BIFTOOL to obtain all starting data allows us to combine this with the manifold computation in a convenient way inside the Matlab environment [13].

### 3.1. Approximating the unstable eigendirection

In order to compute the unstable eigenfunction of the Poincaré operator $D P(q)$ corresponding to a periodic solution $\Gamma$ we consider the linearisation of (1) around $\Gamma$, given by

$$
\begin{equation*}
\frac{\mathrm{d} y(t)}{\mathrm{d} t}=A(t) y(t)+B(t) y(t-\tau) \tag{6}
\end{equation*}
$$

where we have omitted the dependency on $\lambda$ (which is considered to be fixed in this section). The periodic coefficients $A, B$ are defined as (writing $F \equiv F\left(x_{1}, x_{2}, \lambda\right)$ )

$$
\begin{equation*}
A(t)=\left.\frac{\partial F}{\partial x_{1}}\right|_{(\Gamma(t), \Gamma(t-\tau), \lambda)} \quad \text { and } \quad B(t)=\left.\frac{\partial F}{\partial x_{2}}\right|_{(\Gamma(t), \Gamma(t-\tau), \lambda)} . \tag{7}
\end{equation*}
$$

The evolution operator of the linearisation (6) is given by

$$
\begin{equation*}
D \Phi^{t}: \mathscr{C} \rightarrow \mathscr{C} . \tag{8}
\end{equation*}
$$

The eigenvalues of $D \Phi^{\top}$, where T is the period of $\Gamma$, are the Floquet multipliers $\mu$ of the periodic solution. The operator $D \Phi^{\top}$ has only a point spectrum (except possibly for the eigenvalue zero). Hence, for each $\mu \neq 0$ there exists a corresponding eigenfunction segment $q_{\mu} \in \mathscr{C}$, such that,

$$
\begin{equation*}
D \Phi^{\top} q_{\mu}=\mu q_{\mu} \tag{9}
\end{equation*}
$$

Moreover, it follows that $q_{\mu}$ is an eigenfunction segment of a trajectory $y_{\mu}$ which repeats itself after time T scaled by the factor $\mu$,

$$
\begin{equation*}
y_{\mu}(t+\mathrm{T})=\mu y_{\mu}(t) \quad \text { for all } t \tag{10}
\end{equation*}
$$

We briefly sketch how the flow (6), and correspondingly the operator $D \Phi^{\top}$, is approximated within the package DDE-BIFTOOL; see [6] for details. The periodic solution $\Gamma$ itself is approximated by a piecewise polynomial solution (of Lagrange polynomials) over a finite mesh. The DDE is satisfied at the so-called collocation points, yielding collocation equations for the coefficients of the polynomials. The collocation equations are used to obtain a discrete mapping (a matrix $M$ ) of the approximation on the function segment $[-\tau / T, 0]$ to the approximation on the function segment $[1-\tau / T, 1]$ (notice the scaling of time). The eigenvalues of this matrix $M$ are then computed and they approximate the Floquet multipliers; the convergence of the multiplier approximations was investigated in [24]. Likewise, the eigenvectors of $M$ are discrete approximations of the eigenfunction segments $q_{\mu}$ on $[-\tau, 0]$. Once an approximation is known on $[-\tau / T, 0]$, these eigenfunction segments are expanded using the collocation equations. In this way, we obtain an approximation of $y_{\mu}$ on $[-\mathrm{T}, 0]$. Except for the time dependency in (6), this scheme is completely equivalent to the collocation equations used for obtaining the periodic solution $\Gamma$ itself. Note that this eigenfunction approximation is not periodic on $[-\mathrm{T}, 0]$, a fact that we will revisit in Section 3.2.

Since the collocation mesh used in (6) for obtaining the approximate eigendirections is exactly the same as the one used for obtaining $\Gamma$, approximations of $\Gamma$ at the mesh points used in the discretisation of (6) are readily available. We remark that refining the mesh for computing the eigendirection compared to the mesh for obtaining $\Gamma$ does not appear to be sensible, because the overall accuracy is limited in any case by the accuracy of the approximation obtained for $\Gamma$.

The above scheme is now fully implemented as part of DDE-BIFTOOL. For a given Floquet multiplier $\mu$ the user has access to the eigenfunction $y_{\mu}$ over the period interval $[-\mathrm{T}, 0]$ of $\Gamma$. This is useful in several situations, for example, when switching between branches of periodic orbits at bifurcation points and indeed to generate starting data for 1D unstable manifold computations. We note that this approach calculates all eigenfunctions corresponding to all Floquet multipliers with real part greater than a user defined value. However, in the case of computing 1D unstable manifolds we only consider one unstable Floquet multiplier and its associated eigenfunction.

### 3.2. Constructing $q_{\delta}^{ \pm}$

We find starting data for the computation of 1D unstable manifolds of a saddle periodic orbit $\Gamma$ with one unstable Floquet multiplier as follows. First, DDE-BIFTOOL is used to obtain data for $\Gamma$ such that it is given as a function over the period interval $[-\mathrm{T}, 0]$. It is possible to ensure that the headpoint $\Gamma(0)$ lies in the section $\Sigma$ by specifying this as an extra condition when computing and correcting the orbit with DDEBIFTOOL. (Note that this is slightly different from the approach taken in [20] where interpolation was used to obtain an orbit $\Gamma$ with head point in $\Sigma$.) We then use the new functionality of DDE-BIFTOOL to find the unstable eigentrajectory $y_{\mu}$ as detailed in the previous section. $\Gamma$ and $y_{\mu}$ represented in this way constitute the 'raw data' we need.

While the arguments that follow are valid for any generic physical space $X$ and section $\Sigma$, we illustrate our method with specific data in Figs. 1 and 2 of the PCF laser introduced in more detail in Section 4 below. For orientation we already mention that this system has the three-dimensional $(E, N)$-space as its physical space, where $E=E_{x}+\mathrm{i} E_{y}$ is a complex variable and $N$ is real. The section is chosen as $\Sigma \equiv\left\{(E, N) \mid E_{x}=0\right\}$.

Fig. 1 shows the 'raw data', namely a saddle periodic orbit $\Gamma$ of the PCF laser with one unstable Floquet multiplier $\mu$, together with the associated unstable eigendirection $E^{u}(\Gamma)$ given by $\left(\Gamma+\delta y_{\mu}\right)(t)$. (In the figure we use the value of $\delta=3.0$ and plot the three components $E_{x}, E_{y}$ and $N$ separately.) The length of the vector
$\left(\Gamma+\delta y_{\mu}\right)(0)$ at $\Gamma(0)$ is approximately $\mu=3.14897$ times that of $\left(\Gamma+\delta y_{\mu}\right)(-\mathrm{T})$ at $\Gamma(-\mathrm{T})$, illustrating that the Floquet multiplier $\mu$ measures the linear expansion over one period T. Notice that $\Gamma_{E_{x}}(0)=0$, which means that $\Gamma$ indeed lies in the section $\Sigma$.

Fig. 2(a) and (b) show the same 'raw data' as Fig. 1, but now plotted in two projections onto the physical space; namely $(E, N)$-space and the $E$-plane, respectively. Similarly, Fig. 2(c) and (d) show the respective data $\Gamma-\delta y_{\mu}$ for $\delta=3.0$, that is, the 'other side' of $E^{u}(\Gamma)$. Also plotted in bold are the points $q_{\delta}^{+}$and $q_{\delta}^{-}$that we are seeking, whose headpoints lie in $\Sigma$. It is immediately apparent from this figure, that

$$
\begin{equation*}
q_{\delta}^{ \pm} \neq q \pm \delta q_{\mu} \tag{11}
\end{equation*}
$$

where $q$ is the fixed point associated with $\Gamma$ and $q_{\mu}$ the respective eigenfunction, both over the interval $[-\tau, 0]$.

The desired points $q_{\delta}^{+}$and $q_{\delta}^{-}$can be extracted from the raw data as follows. In the first case, sketched in Fig. 2(a) and (b), the vector $y_{\mu}(0)$ points in the same direction as the flow. This means that the point $\left(\Gamma+\delta y_{\mu}\right)(0)$ lies past the section $\Sigma$. In this situation, the desired point $q_{\delta}^{+}$can be found by moving backwards (by decreasing $t$ ) along the data $\Gamma+\delta y_{\mu}$ and finding the (first) intersection point of this data with $\Sigma$ by interpolation. This point is by definition $q_{\delta}^{+}(0)$ and the data segment of length $\tau$ before this point is $q_{\delta}^{+}$.

The second case is shown in Fig. 2(c) and (d). Here $\left(\Gamma-\delta y_{\mu}\right)(0)$ points in the opposite direction to the flow, so that $\left(\Gamma-\delta y_{\mu}\right)(0)$ lies before the section $\Sigma$. In this situation, the desired point $q_{\delta}^{-}$must be found by moving forward, that is, by increasing $t$. The problem is that $\Gamma-\delta y_{\mu}$ is only known in the time interval $[-\mathrm{T}, 0]$. To obtain data for $t>0$ we extend the data for $\Gamma$ simply by periodicity. However, the data for $y_{\mu}$ is not periodic, owing to the expansion given by the Floquet multiplier $\mu$. Nevertheless, we can extend the data for $y_{\mu}$ by periodicity if we multiply by $\mu$ at the same time, hence, using the scaling (10). Once the data has been extended in this way, we move forward in the data, find the intersection point $q_{\delta}^{-}(0)$ with $\Sigma$ by interpolation and obtain $q_{\delta}^{-}$as the data segment of length $\tau$ before this point.

This procedure is now fully implemented in Matlab [12], so that it is integrated with the DDE-BIFTOOL routines. As output we write out both $q$ and $q_{\delta}^{ \pm}$for a $\delta$ that was specified by the user. This output forms the starting data as needed by the 1D manifold algorithm introduced in [20].

The fact that the vector $q_{\mu}(0)$ generically points out of the section $\Sigma$ has the consequence that the intersection of the linear space $E^{u}(\Gamma)$ with the plane $\Sigma$ is not a straight line. The two branches of this intersection curve are given by the parametrised families $\left\{q_{\delta}^{ \pm}(0) \mid \delta \in \mathbb{R}^{\geqslant 0}\right\}$, which is shown in Fig. 3. Also shown is part of the global 1D unstable manifold $W^{u}$ computed using our algorithm with an initial distance


Fig. 3. Unstable linear eigenspace $E^{u}(q)$ as parametrised by $\left\{q_{\delta}^{ \pm}(0) \mid \delta \in \mathbb{R}^{\geqslant}\right\}$, together with the unstable manifold $W^{u}(q)$ as computed with our algorithm (for an initial distance of $\delta=0.001$ ); individual dots are for the stated values of $\delta$.
of $\delta=0.001$. It is clear that $E^{u}(\Gamma) \cap \Sigma$ is indeed not a straight line, but a curve that is tangent to $W^{u}$ at $q$. (Note that at $q, E^{u}(q)$ is tangent to $W^{u}(q)$ not only at $q(0)$ but also along the entire history of $q$.) Notice that, for large values of $\delta$, the linear space $E^{u}(\Gamma)$ is a very bad approximation to the 1 D unstable manifold $W^{u}$; see Fig. 3(b). It is therefore very important to choose $\delta$ small enough when starting a manifold computation. In Figs. 1 and 2 we chose $\delta=3.0$ purely for illustrative purposes. Fig. 3(b) shows that an initial distance of $\delta=3.0$ along $E^{u}(\Gamma)$ would give very bad starting data for computing the manifold $W^{u}(\Gamma)$.

## 4. Example: Phase-conjugate feedback laser

For illustration we consider the technologically relevant example of a DDE describing a PCF laser [ $9,10,19]$. This system can be modelled by the DDEs

$$
\begin{align*}
\frac{\mathrm{d} E(t)}{\mathrm{d} t} & =\frac{1}{2}\left[-\mathrm{i} \alpha G_{N}\left(N(t)-N_{\mathrm{sol}}\right)+\left(G(t)-\frac{1}{\tau_{p}}\right)\right] E(t)+\kappa E^{*}(t-\tau),  \tag{12}\\
\frac{\mathrm{d} N(t)}{\mathrm{d} t} & =\frac{I}{q}-\frac{N(t)}{\tau_{e}}-G(t)|E(t)|^{2},
\end{align*}
$$

describing the evolution of the complex electric field $E(t)=E_{x}(t)+\mathrm{i} E_{y}(t)$ and the population inversion $N(t)$. Nonlinear gain is included in the term $G(t)=G_{N}\left(N(t)-N_{0}\right)(1-\epsilon P(t))$ where $\epsilon=3.57 \times 10^{-8}$ is the nonlinear gain coefficient and $P(t)=|E(t)|^{2}$ is the intensity. All other parameters are set to realistic values corresponding to a $\mathrm{Ga}-\mathrm{Al}-\mathrm{As}$ semiconductor laser and are given in [13-16,19,20]. The phase-conjugate feedback term involves the feedback rate $\kappa$ and the external cavity round-trip time $\tau$, which we fix at the value of $\tau=2 / 3 \mathrm{~ns}$, corresponding to an external cavity length $L_{\text {ext }} \approx 10 \mathrm{~cm}$. We consider changes in the dynamics of the PCF laser as the dimensionless parameter $\kappa \tau$ is varied. Finally, we note that (12) has $\mathbb{Z}_{2}-$ symmetry given by the transformation $(E, N) \rightarrow(-E, N)$. Consequently, every invariant set is either symmetric or has a symmetric counterpart under this symmetry.

### 4.1. General dynamics of the PCF laser

In [19] it was shown that the general picture of the dynamics of the PCF laser is that of stable periodic operation interspersed with 'bubbles' of more complicated, for the most part, chaotic dynamics. One refers to these stable periodic orbits as external cavity modes (ECMs) of the PCF laser [13]. The bubbles of chaotic dynamics are the result of competition between ECMs.

The transitions between ECMs and the regions of chaos may be very sudden. One such transition was studied in [15] where the transition at the start of the second bubble of chaos, from a locked periodic solution to a chaotic solution, was shown to be due to the break-up of a torus culminating in a crisis bifurcation. This study employed manifold computations to identify the shape of the underlying torus as one passes through the region of locking before the crisis bifurcation and the ensuing chaotic dynamics. (Note that the iterative power method worked for all examples in [20], meaning that we were able to compute both branches of the unstable manifold in each case.)

The bifurcation diagram in Fig. 4(a) shows the first bubble of chaos in the PCF laser. It was obtained by simulation, where we plot, after transients have died down, a normalised inversion $\hat{N}$ whenever the power $P(t)=|E(t)|^{2}$ crosses its average value; compare [15,19]. Fig. 4(a) reveals a steady state locked solution for $\kappa \tau<0.75$. Physically this corresponds to a frequency match between the solitary laser and the phaseconjugating mirror; see $[13,14,16]$ for detailed studies of this locked solution. A periodic solution then appears which is seen to undergo a period-doubling route to the first bubble of chaos. One then observes a sudden transition to the first ECM solution at $\kappa \tau \approx 1.86$.


Fig. 4. Bifurcation diagrams of the PCF laser obtained by simulation.

An enlargement of the bifurcation diagram near this transition is shown in Fig. 4(b). Note that there is no hysteresis in this transition, so that when decreasing $\kappa \tau$ one sees a sudden jump from periodic output to chaos. Fig. 4(b) reveals that the curve of stable periodic solutions has a quadratic tangency with the chaotic region, which is typical for a saddle-node bifurcation of periodic orbits. This phenomenon was already found in [19] at the end of the second bubble of chaos, where it was identified as an intermittent transition (or a saddle-node bifurcation of periodic orbits that takes place on a chaotic attractor). The saddle-node bifurcation itself was found in [13] by a continuation study of the periodic solutions of the PCF laser.

### 4.2. Intermittent transition to chaos

Here we use manifold computations to show that the sudden transition to periodic operation at the end of the first bubble of chaos is indeed due to an intermittent transition. Specifically, we compute the unstable manifold associated with the saddle periodic solution born in the saddle-node bifurcation of periodic orbits. Due to the closeness of the bifurcation one needs good starting data. In fact, the examples presented in this paper could not be computed with the iterative power method, that is, we could only find one of the branches of the unstable manifold. However, the method presented in Section 3 does indeed allow us to compute both branches of all relevant manifolds.

Fig. 5 shows the trace in the section $\Sigma$ defined by $E_{x}=0$ of the unstable manifold $W^{u}(q)$ of the saddle fixed point $q$ for four values of $\kappa \tau$ as one moves away from the saddle-node bifurcation at $\kappa \tau \approx 1.86027$. The horizontal cross $(+)$ indicates the point $q(0)$, while the intersection of the bifurcating stable periodic solution with $\Sigma$ is marked by a diagonal cross $(\times)$. In each computation, an initial step $q_{\delta}^{ \pm}$for $\delta=0.001$ along $E^{u}(q)$ was used. Other accuracy parameters, detailed in [20], were set to $\alpha_{\text {min }}=0.2, \alpha_{\text {max }}=0.3$, $(\Delta \alpha)_{\min }=5.0 \times 10^{-4},(\Delta \alpha)_{\max }=5.0 \times 10^{-3}, \Delta_{\min }=5.0 \times 10^{-2}$ and $\varepsilon=0.2$.

Fig. 5(a) shows the unstable manifold for $\kappa \tau \approx 1.86036$, closest to the saddle-node bifurcation. Here we see that the short branch converges quickly to the stable periodic solution $(\times)$. The other branch, on the other hand, is very long and makes very large excursions, where it follows the previous chaotic attractor. To illustrate this, we show the intersection of this chaotic attractor with $\Sigma$ as a cloud of grey dots. (Note that for the used values of $\kappa \tau$, past the bifurcation at $\kappa \tau \approx 1.86027$, the chaotic attractor does not exist any longer; it is plotted strictly for illustration and is the same in all panels of Fig. 5.) As one moves away from the transition, the short branch grows in length but always converges to the stable periodic solution, as is expected near a saddle-node bifurcation. The long branch does not change so much, and principally follows
the old chaotic attractor; see Fig. 5(b)-(d). This shows that we are indeed dealing with an intermittent transition: as one moves away from the bifurcation point, the saddle periodic orbit and the stable periodic orbit get further apart; while at the transition the periodic orbits disappear and the long branch of the unstable manifold forms the chaotic attractor.

The reader will notice a 'gap' in the upper right of the unstable manifold shown in Fig. 5(a)-(c). This gap shortens as $\kappa \tau$ is increased and finally closes in Fig. 5(d). As was described in [20], it is due to the orbit of the flow becoming tangent to the section $\Sigma$. The tangencies are detected by monitoring the integration time used in computing previous points on the manifold. If this tangency occurs at an interior point of the flow, our algorithm changes the number of returns of the Poincare map to $\Sigma$ and the computation continues. If the tangency occurs at the end point of the orbit of the flow an integration time constraint causes the manifold computation to stop. The gaps in the manifolds of Fig. 5(a)-(c) are due to a tangency at an end point. However, by specifying a large enough time constraint, the algorithm continues to search for new images in $\Sigma$ of the previously computed part of the manifold and resumes computing the manifold. If this is the case, condition (5) is ignored, resulting in a gap in the manifold. In this situation it is possible that the


Fig. 5. Both branches of the 1D unstable manifold of the saddle point $(+)$, the short branch converges to the attractor $(\times)$, and the other branch to the chaotic attractor (grey) that exists before the crisis. From (a) to (d) $\kappa \tau$ takes the values $1.86036,1.86238,1.87774$ and 1.89159 .
algorithm may have switched to another branch of the manifold. Therefore, it is important to check that the correct branch is being followed. We do this in two ways. Firstly, we check that the discontinuous branches indeed match up exactly, as in Fig. 5(d), as the parameter is changed. Secondly, we monitor the orbit itself in phase-space and detect its tangency. In the computations of Fig. 5 we allowed for a difference in integration times between consecutive mesh points on the manifold of at most $20 \%$.

## 5. Conclusions

We used the continuation package DDE-BIFTOOL to compute unstable eigenfunctions of saddle periodic orbits. We believe this to be the first time that this was done for DDEs. From these eigenfunctions we then constructed reliable starting data for 1D unstable manifold computations. This allows us to perform manifold computations in situations where the iterative power method we used previously in $[15,20]$ ran into difficulty. As an example, we investigated a sudden transition to chaos in a semiconductor laser subject to phase-conjugate feedback and were able to show that this transition was due to a saddle-node bifurcation of limit cycles taking place on the chaotic attractor, known as an intermittent transition.

We expect that the study of global dynamics and bifurcations in DDEs arising in applications will be a topic of much interest in the coming years. The algorithm for computing 1D unstable manifolds constitutes a valuable new tool. It is now integrated with DDE-BIFTOOL inside the Matlab environment [12].

The approach taken here to generate starting data can be generalised to the case of more unstable eigenfunctions. The next challenging step is to generalise the algorithm in [23] to compute 2D unstable manifolds of equilibria and saddle periodic orbits in DDEs.

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