## Lecture Notes on

## NUMERICAL ANALYSIS

of
OF NONLINEAR EQUATIONS

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## Persistence of Solutions

We discuss the persistence of solutions to nonlinear equations.

- Newton's method for solving a nonlinear equation

$$
\mathbf{G}(\mathbf{u})=\mathbf{0}, \quad \mathbf{G}(\cdot), \mathbf{u} \in \mathrm{R}^{n}
$$

may not converge if the "initial guess" is not close to a solution.

- To alleviate this problem one can introduce an artificial "homotopy" parameter in the equation.
- Actually, most equations already have parameters.
- We discuss the persistence of solutions to such parameter-dependent equations.


## The Implicit Function Theorem

Let $\mathbf{G}: \mathrm{R}^{n} \times \mathrm{R} \rightarrow \mathrm{R}^{n}$ satisfy
(i) $\quad \mathbf{G}\left(\mathbf{u}_{0}, \lambda_{0}\right)=\mathbf{0}, \quad \mathbf{u}_{0} \in \mathrm{R}^{n}, \quad \lambda_{0} \in \mathrm{R}$.
(ii) $\mathbf{G}_{\mathbf{u}}\left(\mathbf{u}_{0}, \lambda_{0}\right)$ is nonsingular (i.e., $\mathbf{u}_{0}$ is an isolated solution),
(iii) $\mathbf{G}$ and $\mathbf{G}_{\mathbf{u}}$ are smooth near $\mathbf{u}_{0}$.

Then there exists a unique, smooth solution family $\mathbf{u}(\lambda)$ such that

- $\mathbf{G}(\mathbf{u}(\lambda), \lambda)=\mathbf{0}, \quad$ for all $\lambda$ near $\lambda_{0}$,
- $\mathbf{u}\left(\lambda_{0}\right)=\mathbf{u}_{0}$.

PROOF : See a good Analysis book ...

EXAMPLE: (AUTO demo hom.)

Let

$$
g(u, \lambda)=\left(u^{2}-1\right)\left(u^{2}-4\right)+\lambda u^{2} e^{c u}
$$

where $c$ is fixed, e.g., $c=0.1$.

When $\lambda=0$ the equation

$$
g(u, 0)=0
$$

has four solutions, namely,

$$
u= \pm 1, \quad \text { and } \quad u= \pm 2
$$

We have

$$
\left.\left.g_{u}(u, \lambda)\right|_{\lambda=0} \equiv \frac{d}{d u}(u, \lambda)\right|_{\lambda=0} \quad=\quad 4 u^{3}-10 u
$$

Since

$$
g_{u}(u, 0)=4 u^{3}-10 u
$$

we have

$$
\begin{array}{ll}
g_{u}(-1,0)=6, & g_{u}(1,0)=-6, \\
g_{u}(-2,0)=-12, & g_{u}(2,0)=12,
\end{array}
$$

which are all nonzero.

Hence each of the four solutions when $\lambda=0$ is isolated .

Thus each of these solutions persists as $\lambda$ becomes nonzero, ( at least for "small" values of $|\lambda|$ ).


Four solution families of $g(u, \lambda)=0$. Note the fold.

- Each of the four solutions at $\lambda=0$ is isolated.
- Thus each of these solutions persists as $\lambda$ becomes nonzero.
- Only two of the four homotopies "reach" $\lambda=1$.
- The two other homotopies meet at a fold.
- IFT condition (ii) is not satisfied at this fold. (Why not?)

Consider the equation

$$
\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0}, \quad \mathbf{u}, \mathbf{G}(\cdot, \cdot) \in \mathrm{R}^{n}, \quad \lambda \in \mathrm{R}
$$

Let

$$
\mathbf{x} \equiv(\mathbf{u}, \lambda)
$$

Then the equation can be written

$$
\mathbf{G}(\mathbf{x})=\mathbf{0}, \quad \mathbf{G}: \mathrm{R}^{n+1} \rightarrow \mathrm{R}^{n}
$$

## DEFINITION.

A solution $\mathbf{x}_{0}$ of $\mathbf{G}(\mathbf{x})=\mathbf{0}$ is regular if the matrix

$$
\mathbf{G}_{\mathbf{x}}^{0} \equiv \mathbf{G}_{\mathbf{x}}\left(\mathbf{x}_{0}\right), \quad(\text { with } n \text { rows and } n+1 \text { columns })
$$

has maximal rank, i.e., if

$$
\operatorname{Rank}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=n
$$

In the parameter formulation,

$$
\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0},
$$

we have

$$
\operatorname{Rank}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Rank}\left(\mathbf{G}_{\mathbf{u}}^{0} \mid \mathbf{G}_{\lambda}^{0}\right)=n \Longleftrightarrow\left\{\begin{array}{l}
\text { (i) } \mathbf{G}_{\mathbf{u}}^{0} \text { is nonsingular, } \\
\text { or } \\
\text { (ii) }\left\{\begin{array}{l}
\operatorname{dim} \mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=1, \\
\text { and } \\
\mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right) .
\end{array}\right.
\end{array}\right.
$$

Above,

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right) \text { denotes the null space of } \mathbf{G}_{\mathbf{u}}^{0},
$$

and

$$
\mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right) \text { denotes the range of } \mathbf{G}_{\mathbf{u}}^{0} \text {, }
$$

i.e., the linear space spanned by the $n$ columns of $\mathbf{G}_{\mathbf{u}}^{0}$.

THEOREM. Let

$$
\mathbf{x}_{0} \equiv\left(\mathbf{u}_{0}, \lambda_{0}\right)
$$

be a regular solution of

$$
\mathrm{G}(\mathrm{x})=0
$$

Then, near $\mathbf{x}_{0}$, there exists a unique one-dimensional solution family

$$
\mathbf{x}(s) \quad \text { with } \quad \mathbf{x}(0)=\mathbf{x}_{0} .
$$

PROOF. Since

$$
\operatorname{Rank}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Rank}\left(\mathbf{G}_{\mathbf{u}}^{0} \mid \mathbf{G}_{\lambda}^{0}\right)=n,
$$

then either $G_{u}^{0}$ is nonsingular and by the IFT we have

$$
\mathbf{u}=\mathbf{u}(\lambda) \quad \text { near } \quad \mathbf{x}_{0},
$$

or else we can interchange colums in the Jacobian $\mathbf{G}_{\mathrm{x}}^{0}$ to see that the solution can locally be parametrized by one of the components of $\mathbf{u}$.

Thus a unique solution family passes through a regular solution.

## NOTE:

- Such a solution family is sometimes also called a solution branch .
- Case (ii) above is that of a simple fold, to be discussed later.
- Thus even near a simple fold there is a unique solution family.
- However, near such a fold, the family can not be parametrized by $\lambda$.

EXAMPLE: The $A \rightarrow B \rightarrow C$ reaction. (AUTO demo abc.)

The equations are

$$
\begin{aligned}
u_{1}^{\prime} & =-u_{1}+D\left(1-u_{1}\right) e^{u_{3}} \\
u_{2}^{\prime} & =-u_{2}+D\left(1-u_{1}\right) e^{u_{3}}-D \sigma u_{2} e^{u_{3}} \\
u_{3}^{\prime} & =-u_{3}-\beta u_{3}+D B\left(1-u_{1}\right) e^{u_{3}}+D B \alpha \sigma u_{2} e^{u_{3}}
\end{aligned}
$$

where
$1-u_{1}$ is the concentration of $A, \quad u_{2}$ is the concentration of $B$,
$u_{3}$ is the temperature, $\alpha=1, \quad \sigma=0.04, \quad B=8$,
$D$ is the Damkohler number, $\quad \beta=1.21$ is the heat transfer coefficient .

We compute stationary solutions for varying $D$.


Stationary solution families of the $A \rightarrow B \rightarrow C$ reaction for $\beta=1.15$.
Solid/dashed lines denote stable/unstable solutions.
The red square denotes a Hopf bifurcation.
(AUTO demo abc). Note the two folds.

## Examples of IFT Application

Here we give examples where the IFT shows that a given solution persists, at least locally, when a problem parameter is changed. We also identify some cases where the conditions of the IFT are not satisfied.

## A Predator-Prey Model

(AUTO demo pp2.)

$$
\left\{\begin{array}{l}
u_{1}^{\prime}=3 u_{1}\left(1-u_{1}\right)-u_{1} u_{2}-\lambda\left(1-e^{-5 u_{1}}\right) \\
u_{2}^{\prime}=-u_{2}+3 u_{1} u_{2}
\end{array}\right.
$$

Here $u_{1}$ may be thought of as "fish" and $u_{2}$ as "sharks", while the term

$$
\lambda\left(1-e^{-5 u_{1}}\right)
$$

represents "fishing", with "fishing-quota" $\lambda$.

When $\lambda=0$ the stationary solutions are

$$
\left.\begin{array}{ll}
3 u_{1}\left(1-u_{1}\right)-u_{1} u_{2} & =0 \\
-u_{2}+3 u_{1} u_{2} & =0
\end{array}\right\} \Rightarrow\left(u_{1}, u_{2}\right)=(0,0),(1,0),\left(\frac{1}{3}, 2\right)
$$

The Jacobian matrix is

$$
\begin{gathered}
\mathbf{G}_{\mathbf{u}}=\left(\begin{array}{ll}
3-6 u_{1}-u_{2}-5 \lambda e^{-5 u_{1}} & -u_{1} \\
3 u_{2} & -1+3 u_{1}
\end{array}\right)=\mathbf{G}_{\mathbf{u}}\left(u_{1}, u_{2} ; \lambda\right) \\
\mathbf{G}_{\mathbf{u}}(0,0 ; 0)=\left(\begin{array}{rr}
3 & 0 \\
0 & -1
\end{array}\right) ; \text { eigenvalues } 3,-1 \quad \text { (unstable) } \\
\mathbf{G}_{\mathbf{u}}(1,0 ; 0)=\left(\begin{array}{rr}
-3 & -1 \\
0 & 2
\end{array}\right) ; \text { eigenvalues }-3,2 \quad \text { (unstable) } \\
\mathbf{G}_{\mathbf{u}}\left(\frac{1}{3}, 2 ; 0\right)=\left(\begin{array}{rr}
-1 & -\frac{1}{3} \\
6 & 0
\end{array}\right) ; \text { eigenvalues }\left\{\begin{array}{l}
(-1-\mu)(-\mu)+2=0 \\
\mu^{2}+\mu+2=0 \\
\mu_{ \pm}=\frac{-1 \pm \sqrt{-7}}{2} \\
\operatorname{Re}\left(\mu_{ \pm}\right)<0
\end{array}\right. \text { (stable). }
\end{gathered}
$$

All three Jacobians at $\lambda=0$ are nonsingular.
Thus, by the IFT, all three stationary points persist for (small) $\lambda \neq 0$.

In this problem we can explicitly find all solutions (see Figure 1) :
Branch I:

$$
\left(u_{1}, u_{2}\right)=(0,0) .
$$

Branch II :

$$
\begin{gathered}
u_{2}=0, \quad \lambda=\frac{3 u_{1}\left(1-u_{1}\right)}{1-e^{-5 u_{1}}} . \\
\text { (Note that } \lim _{u_{1} \rightarrow 0} \lambda=\lim _{u_{1} \longrightarrow 0} \frac{3\left(1-2 u_{1}\right)}{5 e^{-5 u_{1}}}=\frac{3}{5} . \text { ) }
\end{gathered}
$$

Branch III :

$$
u_{1}=\frac{1}{3}, \quad \frac{2}{3}-\frac{1}{3} u_{2}-\lambda\left(1-e^{-5 / 3}\right)=0 \Rightarrow u_{2}=2-3 \lambda\left(1-e^{-5 / 3}\right) .
$$

These solution families intersect at two branch points, one of which is

$$
\left(u_{1}, u_{2}, \lambda\right)=(0,0,3 / 5) .
$$



Figure 1: Stationary solution families of the predator-prey model. Solid/dashed lines denote stable/unstable solutions. Note the fold, the bifurcations (open squares), and the Hopf bifurcation (red square).


Figure 2: Stationary solution families of the predator-prey model, showing fish versus quota. Solid/dashed lines denote stable/unstable solutions.

- Stability of branch I :

$$
\mathbf{G}_{\mathbf{u}}((0,0) ; \lambda)=\left(\begin{array}{cc}
3-5 \lambda & 0 \\
0 & -1
\end{array}\right) ; \quad \text { eigenvalues } 3-5 \lambda,-1
$$

Hence the trivial solution is :

$$
\text { unstable if } \lambda<3 / 5
$$

and

$$
\text { stable if } \lambda>3 / 5
$$

as indicated in Figure 2.

- Stability of branch II :

This family has no stable positive solutions.

- Stability of branch III :

At

$$
\lambda_{H} \approx 0.67,
$$

(the red square in Figure 2) the complex eigenvalues cross the imaginary axis.

This crossing is a Hopf bifurcation, a topic to be discussed later.

Beyond $\lambda_{H}$ there are periodic solutions whose period $T$ increases as $\lambda$ increases. (See Figure 4 for some representative periodic orbits.)

The period becomes infinite at $\lambda=\lambda_{\infty} \approx 0.70$.

This final orbit is called a heteroclinic cycle.


Figure 3: Stationary (blue) and periodic (red) solution families of the predatorprey model.


Figure 4: Some periodic solutions of the predator-prey model. The largest orbits are very close to a heteroclinic cycle.

From Figure 3 we can deduce the solution behavior for (slowly) increasing $\lambda$ :

- Branch III is followed until $\lambda_{H} \approx 0.67$.
- Periodic solutions of increasing period until $\lambda=\lambda_{\infty} \approx 0.70$.
- Collapse to trivial solution (Branch I).


## EXERCISE.

Use AUTO to repeat the numerical calculations (demo pp2) .
Sketch phase plane diagrams for $\lambda=0,0.5,0.68,0.70,0.71$.

## The Gelfand-Bratu Problem

(AUTO demo exp.)

$$
\left\{\begin{array}{l}
u^{\prime \prime}(x)+\lambda e^{u(x)}=0, \quad \forall x \in[0,1] \\
u(0)=u(1)=0
\end{array}\right.
$$

If $\lambda=0$ then $u(x) \equiv 0$ is a solution.

We'll prove that this solution is isolated, so that there is a continuation

$$
u=\tilde{u}(\lambda), \quad \text { for }|\lambda| \text { small } .
$$

Consider

$$
\left.\begin{array}{l}
u^{\prime \prime}(x)+\lambda e^{u(x)}=0, \\
u(0)=0, \quad u^{\prime}(0)=p,
\end{array}\right\} \Rightarrow u=u(x ; p, \lambda)
$$

We want to solve

$$
\underbrace{u(1 ; p, \lambda)}_{\equiv G(p, \lambda)}=0
$$

for $|\lambda|$ small.
Here

$$
G(0,0)=0
$$

We must show (IFT) that

$$
\left.\begin{array}{r}
G_{p}(0,0) \equiv u_{p}(1 ; 0,0) \neq 0: \\
u_{p}^{\prime \prime}(x)+\lambda_{0} e^{u_{0}(x)} u_{p}(x)=0, \\
u_{p}(0)=0, \quad u_{p}^{\prime}(0)=1,
\end{array}\right\} \quad \text { where } \quad u_{0}(x) \equiv 0 .
$$

Now $u_{p}(x ; 0,0)$ satisfies

$$
\left\{\begin{array}{l}
u_{p}^{\prime \prime}(x)=0, \\
u_{p}(0)=0, \quad u_{p}^{\prime}(0)=1
\end{array}\right.
$$

Hence

$$
u_{p}(x ; 0,0)=x, \quad u_{p}(1 ; 0,0)=1 \neq 0
$$

EXERCISE. Compute the solution family of the Gelfand-Bratu problem as represented in Figures 5 and 6. (AUTO demo exp.)


Figure 5: Bifurcation diagram of the Gelfand-Bratu equation.


Figure 6: Some solutions to the Gelfand-Bratu equation.

## A Nonlinear Eigenvalue Problem

(AUTO demo nev.)
Consider the nonlinear boundary value problem

$$
\left\{\begin{array}{c}
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \\
u(0)=u(1)=0
\end{array}\right.
$$

which has $u(t) \equiv 0$ as solution for all $\lambda$.
Equivalently, we want the solution

$$
u=u(t ; p, \lambda)
$$

of the initial value problem

$$
\left\{\begin{array}{l}
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \\
u(0)=0, \quad u^{\prime}(0)=p
\end{array}\right.
$$

that satisfies

$$
G(p, \lambda) \equiv u(1 ; p, \lambda)=0
$$

Here

$$
G: \mathrm{R} \times \mathrm{R} \rightarrow \mathrm{R},
$$

with

$$
G(0, \lambda)=0, \quad \text { for all } \lambda .
$$

Let

$$
\begin{aligned}
u_{p}(t ; p, \lambda) & =\frac{d u}{d p}(t ; p, \lambda), \\
G_{p}(p, \lambda) & =u_{p}(1 ; p, \lambda) .
\end{aligned}
$$

Then $u_{p}\left(=u_{p}^{0}\right)$ satisfies

$$
\left\{\begin{array}{l}
u_{p}^{\prime \prime}+\lambda(1+2 u) u_{p}=0 \\
u_{p}(0)=0, \quad u_{p}^{\prime}(0)=1
\end{array}\right.
$$

which, about $u \equiv 0$, gives

$$
\left\{\begin{array}{l}
u_{p}^{\prime \prime}+\lambda u_{p}=0 \\
u_{p}(0)=0, \quad u_{p}^{\prime}(0)=1
\end{array}\right.
$$

By the variation of parameters formula

$$
\begin{gathered}
u_{p}(t ; p, \lambda)=\frac{\sin \sqrt{\lambda} t}{\sqrt{\lambda}}, \quad \lambda \geq 0, \quad(\text { independent of } p) \\
G_{p}(0, \lambda)=u_{p}(1 ; p, \lambda)=\frac{\sin (\sqrt{\lambda})}{\sqrt{\lambda}}=0, \quad \text { if } \lambda=\lambda_{k} \equiv(k \pi)^{2}
\end{gathered}
$$

Thus the conditions of the IFT fail to be satisfied at $\lambda_{k}=(k \pi)^{2}$. (We will see that these solutions are branch points .)


Figure 7: Solution families to the nonlinear eigenvalue problem.


Figure 8: Some solutions to the nonlinear eigenvalue problem.

## Numerical Continuation

Here we discuss algorithms for computing families of solutions to nonlinear equations. The IFT is important in the design of such continuation methods.

- Newton's method for solving a nonlinear equation

$$
\mathbf{G}(\mathbf{u})=\mathbf{0}, \quad \mathbf{G}(\cdot), \mathbf{u} \in \mathrm{R}^{n}
$$

may not converge if the "initial guess" is not close to a solution.

- To deal with this one can introduce a "homotopy parameter".
- Most equations already naturally have parameters.
- We now discuss computing such families of solutions.

Consider the equation

$$
\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0}, \quad \mathbf{u}, \mathbf{G}(\cdot, \cdot) \in \mathrm{R}^{n}, \quad \lambda \in \mathrm{R}
$$

Let

$$
\mathbf{x} \equiv(\mathbf{u}, \lambda)
$$

Then the equation can be written

$$
\mathbf{G}(\mathbf{x})=\mathbf{0}, \quad \mathbf{G}: \mathrm{R}^{n+1} \rightarrow \mathrm{R}^{n}
$$

## DEFINITION.

A solution $\mathbf{x}_{0}$ of $\mathbf{G}(\mathbf{x})=\mathbf{0}$ is regular if the $n$ (rows) by $n+1$ (columns) matrix

$$
\mathrm{G}_{\mathrm{x}}^{0} \equiv \mathrm{G}_{\mathbf{x}}\left(\mathrm{x}_{0}\right)
$$

has maximal rank, i.e., if

$$
\operatorname{Rank}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=n
$$

In the parameter formulation,

$$
\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0},
$$

we have

$$
\operatorname{Rank}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Rank}\left(\mathbf{G}_{\mathbf{u}}^{0} \mid \mathbf{G}_{\lambda}^{0}\right)=n \Longleftrightarrow\left\{\begin{array}{l}
\text { (i) } \mathbf{G}_{\mathbf{u}}^{0} \text { is nonsingular, } \\
\text { or } \\
\text { (ii) }\left\{\begin{array}{l}
\operatorname{dim} \mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=1, \\
\operatorname{and} \\
\mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right) .
\end{array}\right.
\end{array}\right.
$$

Above,

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right) \text { denotes the null space of } \mathbf{G}_{\mathbf{u}}^{0},
$$

and

$$
\mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right) \text { denotes the range of } \mathbf{G}_{\mathbf{u}}^{0} \text {, }
$$

i.e., the linear space spanned by the $n$ columns of $\mathbf{G}_{\mathbf{u}}^{0}$.

THEOREM. Let

$$
\mathbf{x}_{0} \equiv\left(\mathbf{u}_{0}, \lambda_{0}\right)
$$

be a regular solution of

$$
\mathbf{G}(\mathbf{x})=0
$$

Then, near $\mathbf{x}_{0}$, there exists a unique one-dimensional continuum of solutions

$$
\mathbf{x}(s) \quad \text { with } \quad \mathbf{x}(0)=\mathbf{x}_{0}
$$

PROOF. Since

$$
\operatorname{Rank}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Rank}\left(\mathbf{G}_{\mathbf{u}}^{0} \mid \mathbf{G}_{\lambda}^{0}\right)=n
$$

then either $\mathbf{G}_{\mathbf{u}}^{0}$ is nonsingular and by the IFT we have

$$
\mathbf{u}=\mathbf{u}(\lambda) \quad \text { near } \quad \mathbf{x}_{0}
$$

or else we can interchange colums in the Jacobian $\mathbf{G}_{\mathbf{x}}^{0}$ to see that the solution can locally be parametrized by one of the components of $\mathbf{u}$.

Thus a unique solution family passes through a regular solution.

## NOTE:

- Such a continuum of solutions is called a solution family or a solution branch.
- Case (ii) above, namely,

$$
\operatorname{dim} \mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=1 \quad \text { and } \quad \mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)
$$

is that of a simple fold, to be discussed later.


Figure 9: A solution family with four folds. (AUTO demo abc with $\beta=1.25$.)

## Parameter Continuation

Here the continuation parameter is taken to be $\lambda$.

Suppose we have a solution $\left(\mathbf{u}_{0}, \lambda_{0}\right)$ of

$$
\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0}
$$

as well as the direction vector $\dot{\mathbf{u}}_{0}$.
Here

$$
\dot{\mathbf{u}} \equiv \frac{d \mathbf{u}}{d \lambda}
$$

We want to compute the solution $\mathbf{u}_{1}$ at $\lambda_{1} \equiv \lambda_{0}+\Delta \lambda$.


Figure 10: Graphical interpretation of parameter-continuation.

To solve the equation

$$
\mathbf{G}\left(\mathbf{u}_{1}, \lambda_{1}\right)=\mathbf{0},
$$

for $\mathbf{u}_{1}$ (with $\lambda=\lambda_{1}$ fixed) we use Newton's method

$$
\begin{array}{cl}
\mathbf{G}_{\mathbf{u}}\left(\mathbf{u}_{1}^{(\nu)}, \lambda_{1}\right) \Delta \mathbf{u}_{1}^{(\nu)}=-\mathbf{G}\left(\mathbf{u}_{1}^{(\nu)}, \lambda_{1}\right), & \nu=0,1,2, \cdots . \\
\mathbf{u}_{1}^{(\nu+1)}=\mathbf{u}_{1}^{(\nu)}+\Delta \mathbf{u}_{1}^{(\nu)} . &
\end{array}
$$

As initial approximation use

$$
\mathbf{u}_{1}^{(0)}=\mathbf{u}_{0}+\Delta \lambda \dot{\mathbf{u}}_{0}
$$

If

$$
\mathbf{G}_{\mathbf{u}}\left(\mathbf{u}_{1}, \lambda_{1}\right) \text { is nonsingular, }
$$

and $\Delta \lambda$ sufficiently small, then the Newton convergence theory guarantees that this iteration will converge.

After convergence, the new direction vector $\dot{\mathbf{u}}_{1}$ can be computed by solving

$$
\mathbf{G}_{\mathbf{u}}\left(\mathbf{u}_{1}, \lambda_{1}\right) \dot{\mathbf{u}}_{1}=-\mathbf{G}_{\lambda}\left(\mathbf{u}_{1}, \lambda_{1}\right) .
$$

This equation follows from differentiating

$$
\mathbf{G}(\mathbf{u}(\lambda), \lambda)=\mathbf{0},
$$

with respect to $\lambda$ at $\lambda=\lambda_{1}$.

NOTE:

- $\dot{\mathbf{u}}_{1}$ can be computed without another $L U$-factorization of $\mathbf{G}_{\mathbf{u}}\left(\mathbf{u}_{1}, \lambda_{1}\right)$.
- Thus the extra work to find $\dot{\mathbf{u}}_{1}$ is negligible.

EXAMPLE: The Gelfand-Bratu problem (AUTO demo exp):

$$
u^{\prime \prime}(x)+\lambda e^{u(x)}=0 \quad \text { for } \quad x \in[0,1], \quad u(0)=0, \quad u(1)=0 .
$$

If $\lambda=0$ then $u(x) \equiv 0$ is an isolated solution.
Discretize by introducing a mesh ,

$$
\begin{gathered}
0=x_{0}<x_{1}<\cdots<x_{N}=1 \\
x_{j}-x_{j-1}=h, \quad(1 \leq j \leq N), \quad h=1 / N .
\end{gathered}
$$

The discrete equations are :

$$
\frac{u_{j+1}-2 u_{j}+u_{j-1}}{h^{2}}+\lambda e^{u_{j}}=0, \quad j=1, \cdots, N-1,
$$

with $u_{0}=u_{N}=0$.

Let

$$
\mathbf{u} \equiv\left(\begin{array}{c}
u_{1} \\
u_{2} \\
\cdot \\
u_{N-1}
\end{array}\right)
$$

Then we can write the above as

$$
\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0}
$$

where

$$
\mathrm{G}: \mathrm{R}^{n} \times \mathrm{R} \rightarrow \mathrm{R}^{n}, \quad n \equiv N-1
$$

Parameter-continuation : Suppose we have

$$
\lambda_{0}, \mathbf{u}_{0}, \quad \text { and } \quad \dot{\mathbf{u}}_{0} .
$$

Set

$$
\lambda_{1}=\lambda_{0}+\Delta \lambda .
$$

Newton's method :

$$
\begin{aligned}
\mathbf{G}_{\mathbf{u}}\left(\mathbf{u}_{1}^{(\nu)}, \lambda_{1}\right) \Delta \mathbf{u}_{1}^{(\nu)}=-\mathbf{G}\left(\mathbf{u}_{1}^{(\nu)}, \lambda_{1}\right), & \nu=0,1,2, \cdots, \\
\mathbf{u}_{1}^{(\nu+1)}=\mathbf{u}_{1}^{(\nu)}+\Delta \mathbf{u}_{1}^{(\nu)}, &
\end{aligned}
$$

with

$$
\mathbf{u}_{1}^{(0)}=\mathbf{u}_{0}+\Delta \lambda \dot{\mathbf{u}}_{0} .
$$

After convergence find $\dot{\mathbf{u}}_{1}$ from

$$
\mathbf{G}_{\mathbf{u}}\left(\mathbf{u}_{1}, \lambda_{1}\right) \dot{\mathbf{u}}_{1}=-\mathbf{G}_{\lambda}\left(\mathbf{u}_{1}, \lambda_{1}\right) .
$$

Repeat the above procedure to find $\mathbf{u}_{2}, \mathbf{u}_{3}, \cdots$.

Here

$$
\mathbf{G}_{\mathbf{u}}(\mathbf{u}, \lambda)=\left(\begin{array}{cccccc}
-\frac{2}{h^{2}}+\lambda e^{u_{1}} & \frac{1}{h^{2}} & & & \\
& \frac{1}{h^{2}} & -\frac{2}{h^{2}}+\lambda e^{u_{2}} & \frac{1}{h^{2}} & & \\
& & & \cdot & \cdot & \cdot \\
& & & & & \\
& & & & \frac{1}{h^{2}} & -\frac{2}{h^{2}}+\lambda e^{u_{N-1}}
\end{array}\right)
$$

Thus we must solve a tridiagonal system for each Newton iteration.

The solution family has a fold where the parameter-continuation method fails. (AUTO demo exp: See the earlier Figures 5 and 6).

## Keller's Pseudo-Arclength Continuation

This method allows continuation of a solution family past a fold.

Suppose we have a solution $\left(\mathbf{u}_{0}, \lambda_{0}\right)$ of

$$
\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0}
$$

as well as the direction vector $\left(\dot{\mathbf{u}}_{0}, \dot{\lambda}_{0}\right)$ of the solution branch.

Pseudo-arclength continuation solves the following equations for $\left(\mathbf{u}_{1}, \lambda_{1}\right)$ :

$$
\begin{gathered}
\mathbf{G}\left(\mathbf{u}_{1}, \lambda_{1}\right)=\mathbf{0} \\
\left(\mathbf{u}_{1}-\mathbf{u}_{0}\right)^{*} \dot{\mathbf{u}}_{0}+\left(\lambda_{1}-\lambda_{0}\right) \dot{\lambda}_{0}-\Delta s=0
\end{gathered}
$$

See Figure 11 for a graphical interpretation.


Figure 11: Graphical interpretation of pseudo-arclength continuation.

Solve the equations

$$
\begin{gathered}
\mathbf{G}\left(\mathbf{u}_{1}, \lambda_{1}\right)=\mathbf{0} \\
\left(\mathbf{u}_{1}-\mathbf{u}_{0}\right)^{*} \dot{\mathbf{u}}_{0}+\left(\lambda_{1}-\lambda_{0}\right) \dot{\lambda}_{0}-\Delta s=0
\end{gathered}
$$

for $\left(\mathbf{u}_{1}, \lambda_{1}\right)$ by Newton's method:

$$
\left(\begin{array}{cc}
\left(\mathbf{G}_{\mathbf{u}}^{1}\right)^{(\nu)} & \left(\mathbf{G}_{\lambda}^{1}\right)^{(\nu)} \\
\dot{\mathbf{u}}_{0}^{*} & \dot{\lambda}_{0}
\end{array}\right)\binom{\Delta \mathbf{u}_{1}^{(\nu)}}{\Delta \lambda_{1}^{(\nu)}}=-\binom{\mathbf{G}\left(\mathbf{u}_{1}^{(\nu)}, \lambda_{1}^{(\nu)}\right)}{\left(\mathbf{u}_{1}^{(\nu)}-\mathbf{u}_{0}\right)^{*} \dot{\mathbf{u}}_{0}+\left(\lambda_{1}^{(\nu)}-\lambda_{0}\right) \dot{\lambda}_{0}-\Delta s}
$$

Next direction vector :

$$
\left(\begin{array}{cc}
\mathbf{G}_{\mathbf{u}}^{1} & \mathbf{G}_{\lambda}^{1} \\
\dot{\mathbf{u}}_{0}^{*} & \dot{\lambda}_{0}
\end{array}\right)\binom{\dot{\mathbf{u}}_{1}}{\dot{\lambda}_{1}}=\binom{\mathbf{0}}{1} .
$$

## NOTE:

- In practice $\left(\dot{\mathbf{u}}_{1}, \dot{\lambda}_{1}\right)$ can be computed with one extra backsubstitution.
- The orientation of the branch is preserved if $\Delta s$ is sufficiently small.
- The direction vector must be rescaled, so that indeed $\left\|\dot{\mathbf{u}}_{1}\right\|^{2}+\dot{\lambda}_{1}^{2}=1$.


## THEOREM.

The Jacobian of the pseudo-arclength system is nonsingular at a regular solution point.

PROOF. Let

$$
\mathbf{x} \equiv(\mathbf{u}, \lambda) \in \mathrm{R}^{n+1}
$$

Then pseudo-arclength continuation can be written as

$$
\begin{gathered}
\mathbf{G}\left(\mathbf{x}_{1}\right)=0 \\
\left(\mathbf{x}_{1}-\mathbf{x}_{0}\right)^{*} \dot{\mathbf{x}}_{0}-\Delta s=0, \quad\left(\left\|\dot{\mathbf{x}}_{0}\right\|=1\right) .
\end{gathered}
$$

(See Figure 12 for a graphical interpretation.)

## "X-space"



Figure 12: Parameter-independent pseudo-arclength continuation.

The matrix in Newton's method at $\Delta s=0$ is

$$
\binom{\mathbf{G}_{x}^{0}}{\dot{\mathbf{x}}_{0}^{*}} \text {. }
$$

At a regular solution we have

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Span}\left\{\dot{\mathbf{x}}_{0}\right\} .
$$

We must show that

$$
\binom{\mathbf{G}_{x}^{0}}{\dot{\mathbf{x}}_{0}^{*}}
$$

is nonsingular at a regular solution.

If on the contrary

$$
\binom{\mathbf{G}_{x}^{0}}{\dot{\mathbf{x}}_{0}^{*}}
$$

is singular then

$$
\mathrm{G}_{\mathrm{x}}^{0} \mathbf{z}=0 \quad \text { and } \quad \dot{\mathbf{x}}_{0}^{*} \mathbf{z}=0
$$

for some vector $\mathbf{z} \neq \mathbf{0}$.

Thus

$$
\mathbf{z}=c \dot{\mathbf{x}}_{0}, \quad \text { for some constant } c .
$$

But then

$$
0=\dot{\mathbf{x}}_{0}^{*} \mathbf{z}=c \dot{\mathbf{x}}_{0}^{*} \dot{\mathbf{x}}_{0}=c\left\|\dot{\mathbf{x}}_{0}\right\|^{2}=c,
$$

so that $\mathbf{z}=\mathbf{0}$, which is a contradiction.

## EXAMPLE:

Use pseudo-arclength continuation for the discretized Gelfand-Bratu problem.
Then the matrix

$$
\binom{\mathbf{G}_{\mathbf{x}}}{\dot{\mathbf{x}}^{*}}=\left(\begin{array}{cc}
\mathbf{G}_{\mathbf{u}} & \mathbf{G}_{\lambda} \\
\dot{\mathbf{u}}^{*} & \dot{\lambda}
\end{array}\right),
$$

in Newton's method is a "bordered tridiagonal" matrix :

## Following Folds

When a parameter passes a fold, then the behavior of a system can change drastically. Thus it is useful to determine how the location of a fold changes when a second parameter changes, i.e., we want the compute a "critical stability curve", or a "locus of fold points", in 2-parameter space.

## Simple Folds

A regular solution $\mathbf{x}_{0} \equiv\left(\mathbf{u}_{0}, \lambda_{0}\right)$ of $\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0}$, is called a simple fold if

$$
\operatorname{dim} \mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=1 \quad \text { and } \quad \mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)
$$

From differentiating

$$
\mathbf{G}(\mathbf{u}(s), \lambda(s))=\mathbf{0}
$$

we have

$$
\mathbf{G}_{\mathbf{u}}(\mathbf{u}(s), \lambda(s)) \dot{\mathbf{u}}(s)+\mathbf{G}_{\lambda}(\mathbf{u}(s), \lambda(s)) \dot{\lambda}(s)=\mathbf{0}
$$

In particular,

$$
\mathbf{G}_{\mathbf{u}}^{0} \dot{\mathbf{u}}_{0}=-\dot{\lambda}_{0} \mathbf{G}_{\lambda}^{0}
$$

At a fold we have $\mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)$. Thus

$$
\begin{array}{|l|}
\hline \dot{\lambda}_{0}=0 \\
\hline
\end{array}
$$

Hence $\mathbf{G}_{\mathbf{u}}^{0} \dot{\mathbf{u}}_{0}=0$. Thus, since $\operatorname{dim} \mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=1$, we have

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=\operatorname{Span}\left\{\dot{\mathbf{u}}_{0}\right\}
$$

Differentiating again, we have

$$
\mathbf{G}_{\mathbf{u}}^{0} \ddot{\mathbf{u}}_{0}+\mathbf{G}_{\lambda}^{0} \ddot{\lambda}_{0}+\mathbf{G}_{\mathbf{u u}}^{0} \dot{\mathbf{u}}_{0} \dot{\mathbf{u}}_{0}+2 \mathbf{G}_{u \lambda}^{0} \dot{\mathbf{u}}_{0} \dot{\lambda}_{0}+\mathbf{G}_{\lambda \lambda}^{0} \dot{\lambda}_{0} \dot{\lambda}_{0}=0
$$

At a simple fold $\left(\mathbf{u}_{0}, \lambda_{0}\right)$ let

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=\operatorname{Span}\{\phi\}, \quad\left(\phi=\dot{\mathbf{u}}_{0}\right)
$$

and

$$
\mathcal{N}\left(\left(\mathbf{G}_{\mathbf{u}}^{0}\right)^{*}\right)=\operatorname{Span}\{\boldsymbol{\psi}\}
$$

Multiply by $\boldsymbol{\psi}^{*}$ and use $\dot{\lambda}_{0}=0$ and $\boldsymbol{\psi} \perp \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)$ to find

$$
\psi^{*} \mathbf{G}_{\lambda}^{0} \ddot{\lambda}_{0}+\boldsymbol{\psi}^{*} \mathbf{G}_{\mathrm{uu}}^{0} \phi \boldsymbol{\phi}=0
$$

Here $\boldsymbol{\psi}^{*} \mathbf{G}_{\lambda}^{0} \neq 0$, since $\mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)$. Thus

$$
\ddot{\lambda}_{0}=\frac{-\boldsymbol{\psi}^{*} \mathbf{G}_{\mathbf{u u}}^{0} \phi \phi}{\boldsymbol{\psi}^{*} \mathbf{G}_{\lambda}^{0}}
$$

If the curvature $\ddot{\lambda}_{0} \neq 0$ then $\left(\mathbf{u}_{0}, \lambda_{0}\right)$ is called a simple quadratic fold.

## The Extended System

To continue a fold in two parameters we use the extended system

$$
\begin{aligned}
& \mathbf{G}(\mathbf{u}, \lambda, \mu)=\mathbf{0} \\
& \mathbf{G}_{\mathbf{u}}(\mathbf{u}, \lambda, \mu) \boldsymbol{\phi}=\mathbf{0} \\
& \boldsymbol{\phi}^{*} \boldsymbol{\phi}_{0}-1=0
\end{aligned}
$$

Here $\mu \in \mathrm{R}$ is a second parameter in the equations.
The vector $\phi_{0}$ is from a "reference solution" $\left(\mathbf{u}_{0}, \boldsymbol{\phi}_{0}, \lambda_{0}, \mu_{0}\right)$.
(In practice this is the latest computed solution point on the branch.)
The above system has the form

$$
\mathbf{F}(\mathbf{U}, \mu)=\mathbf{0}, \quad \mathbf{U} \equiv(\mathbf{u}, \phi, \lambda), \quad \mathbf{F}: \mathrm{R}^{2 n+1} \times \mathrm{R} \rightarrow \mathrm{R}^{2 n+1}
$$

or, using the "parameter-free" formulation,

$$
\mathbf{F}(\mathbf{X})=\mathbf{0}, \quad \mathbf{X} \equiv(\mathbf{U}, \mu), \quad \mathbf{F}: \mathrm{R}^{2 n+2} \rightarrow \mathrm{R}^{2 n+1}
$$

## Parameter Continuation

First consider continuing a solution

$$
\left(\mathbf{u}_{0}, \phi_{0}, \lambda_{0}\right) \quad \text { at } \quad \mu=\mu_{0}
$$

in $\mu$ (although, in practice, we use pseudo-arclength continuation).

By the IFT there is a smooth solution family

$$
\mathbf{U}(\mu)=(\mathbf{u}(\mu), \phi(\mu), \lambda(\mu))
$$

if the Jacobian

$$
\mathbf{F}_{\mathbf{U}}^{0} \equiv \frac{d \mathbf{F}}{d \mathbf{U}}\left(\mathbf{U}_{0}\right)=\left(\begin{array}{ccc}
\mathbf{G}_{\mathbf{u}}^{0} & O & \mathbf{G}_{\lambda}^{0} \\
\mathbf{G}_{\mathbf{u u}}^{0} \phi_{0} & \mathbf{G}_{\mathbf{u}}^{0} & \mathbf{G}_{\mathbf{u} \lambda}^{0} \phi_{0} \\
\mathbf{0}^{*} & \boldsymbol{\phi}_{0}^{*} & 0
\end{array}\right)
$$

is nonsingular.

## THEOREM.

A simple quadratic fold with respect to $\lambda$ can be continued locally, using the second parameter $\mu$ as continuation parameter.

PROOF.

Suppose $\mathbf{F}_{\mathbf{U}}^{0}$ is singular. Then

$$
\begin{array}{ll}
\text { (i) } & \mathbf{G}_{\mathbf{u}}^{0} \mathbf{x}+z \mathbf{G}_{\lambda}^{0}=\mathbf{0}, \\
(\text { ii }) & \mathbf{G}_{\mathbf{u u}}^{0} \boldsymbol{\phi}_{0} \mathbf{x}+\mathbf{G}_{\mathbf{u}}^{0} \mathbf{y}+z \mathbf{G}_{\mathbf{u} \lambda}^{0} \boldsymbol{\phi}_{0}=\mathbf{0}, \\
\text { (iii) } & \boldsymbol{\phi}_{0}^{*} \mathbf{y}=0
\end{array}
$$

for some

$$
\mathbf{x}, \mathbf{y} \in \mathrm{R}^{n}, \quad z \in \mathrm{R} .
$$

Since $\mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)$ we have from $(i)$ that $z=0$, and hence

$$
\mathbf{x}=c_{1} \phi_{0}, \quad \text { for some } \quad c_{1} \in \mathrm{R}
$$

Multiply (ii) on the left by $\boldsymbol{\psi}_{0}^{*}$, to get

$$
c_{1} \boldsymbol{\psi}_{0}^{*} \mathbf{G}_{\mathbf{u u}}^{0} \boldsymbol{\phi}_{0} \boldsymbol{\phi}_{0}=0 .
$$

Thus $c_{1}=0$, because by assumption $\boldsymbol{\psi}_{0}^{*} \mathbf{G}_{\mathbf{u u}}^{0} \boldsymbol{\phi}_{0} \boldsymbol{\phi}_{0} \neq 0$.
Therefore $\mathbf{x}=\mathbf{0}$, and from (ii) we now have

$$
\mathbf{G}_{\mathbf{u}}^{0} \mathbf{y}=\mathbf{0}, \quad \text { i.e. }, \quad \mathbf{y}=c_{2} \boldsymbol{\phi}_{0}
$$

But then by (iii) $c_{2}=0$. Thus

$$
\mathbf{x}=\mathbf{y}=\mathbf{0}, \quad z=0
$$

and hence $\mathbf{F}_{\mathbf{U}}^{0}$ is nonsingular.

NOTE:

- The zero eigenvalue of $\mathbf{G}_{\mathbf{u}}^{0}$ need not be be algebraically simple.
- Thus, for example, $\mathbf{G}_{\mathbf{u}}^{0}$ may have the form

$$
\mathbf{G}_{\mathbf{u}}^{0}=\left(\begin{array}{ll}
0 & 1 \\
0 & 0
\end{array}\right)
$$

provided the fold is simple and quadratic.

- Parameter-continuation fails at folds w.r.t $\mu$ on a solution family to

$$
\mathbf{F}(\mathbf{U}, \mu)=\mathbf{0} .
$$

- Such points represent cusps, branch points, or isola formation points.


## Pseudo-Arclength Continuation of Folds

Treat $\mu$ as one of the unknowns, and compute a solution family

$$
\mathbf{X}(s) \equiv(\mathbf{u}(s), \boldsymbol{\phi}(s), \lambda(s), \mu(s))
$$

to

$$
\mathbf{F}(\mathbf{X}) \equiv\left\{\begin{array}{l}
\mathbf{G}(\mathbf{u}, \lambda, \mu)=\mathbf{0}  \tag{1}\\
\mathbf{G}_{\mathbf{u}}(u, \lambda, \mu) \boldsymbol{\phi}=\mathbf{0} \\
\boldsymbol{\phi}^{*} \phi_{0}-1=0
\end{array}\right.
$$

and the added pseudo-arclength equation

$$
\begin{equation*}
\left(\mathbf{u}-\mathbf{u}_{0}\right)^{*} \dot{\mathbf{u}}_{0}+\left(\boldsymbol{\phi}-\boldsymbol{\phi}_{0}\right)^{*} \dot{\phi}_{0}+\left(\lambda-\lambda_{0}\right) \dot{\lambda}_{0}+\left(\mu-\mu_{0}\right) \dot{\mu}_{0}-\Delta s=0 \tag{2}
\end{equation*}
$$

As before,

$$
\left(\dot{\mathbf{u}}_{0}, \dot{\boldsymbol{\phi}}_{0}, \dot{\lambda}_{0}, \dot{\mu}_{0}\right)
$$

is the direction of the branch at the current solution point

$$
\left(\mathbf{u}_{0}, \phi_{0}, \lambda_{0}, \mu_{0}\right)
$$

The Jacobian of $\mathbf{F}$ with respect to $\mathbf{u}, \boldsymbol{\phi}, \lambda$, and $\mu$, at

$$
\mathbf{X}_{0}=\left(\mathbf{u}_{0}, \phi_{0}, \lambda_{0}, \mu_{0}\right),
$$

is now

$$
\mathbf{F}_{\mathbf{X}}^{0} \equiv \frac{d \mathbf{F}}{d \mathbf{X}}\left(\mathbf{X}_{0}\right) \equiv\left(\begin{array}{cccc}
\mathrm{G}_{\mathbf{u}}^{0} & O & \mathrm{G}_{\lambda}^{0} & \mathrm{G}_{\mu}^{0} \\
\mathbf{G}_{\mathbf{u u}}^{0} \boldsymbol{\phi}_{0} & \mathrm{G}_{\mathbf{u}}^{0} & \mathbf{G}_{\mathbf{u}}^{0} \boldsymbol{\phi}_{0} & \mathrm{G}_{\mathbf{u} \mu}^{0} \boldsymbol{\phi}_{0} \\
\mathbf{0}^{*} & \phi_{0}^{*} & 0 & 0
\end{array}\right) .
$$

For pseudo-arclength continuation we must check that $\mathbf{F}_{\mathbf{X}}^{0}$ has full rank.
For a simple quadratic fold with respect to $\lambda$ this follows from the Theorem.
Otherwise, if

$$
\psi_{0}^{*} \mathrm{G}_{\mathrm{uu}}^{0} \phi_{0} \phi_{0} \neq 0,
$$

and

$$
\mathbf{G}_{\lambda}^{0} \in \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right), \quad \mathbf{G}_{\mu}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)
$$

i.e., if we have a simple quadratic fold with respect to $\mu$, then we can apply the theorem to $\mathbf{F}_{\mathbf{X}}^{0}$ with the second last column struck out, to see that $\mathbf{F}_{\mathbf{X}}^{0}$ still has full rank.

EXAMPLE: The $A \rightarrow B \rightarrow C$ reaction. (AUTO demo abc.)

The equations are

$$
\begin{aligned}
u_{1}^{\prime} & =-u_{1}+D\left(1-u_{1}\right) e^{u_{3}} \\
u_{2}^{\prime} & =-u_{2}+D\left(1-u_{1}\right) e^{u_{3}}-D \sigma u_{2} e^{u_{3}} \\
u_{3}^{\prime} & =-u_{3}-\beta u_{3}+D B\left(1-u_{1}\right) e^{u_{3}}+D B \alpha \sigma u_{2} e^{u_{3}},
\end{aligned}
$$

where
$1-u_{1}$ is the concentration of $A, \quad u_{2}$ is the concentration of $B$,
$u_{3}$ is the temperature, $\alpha=1, \quad \sigma=0.04, B=8$,
$D$ is the Damkohler number,
$\beta$ is the heat transfer coefficient.

We will compute solutions for varying $D$ and $\beta$.


Figure 13: A stationary solution family of demo abc; $\beta=1.15$.


Figure 14: The locus of folds of demo abc.


Figure 15: Stationary solution families for $\beta=1.15,1.17, \ldots, 1.39$.

## Numerical Treatment of Bifurcations

Here we discuss branch switching, and the detection of branch points.

## Simple Singular Points

Let

$$
\mathrm{G}: \mathrm{R}^{n+1} \rightarrow \mathrm{R}^{n}
$$

A solution

$$
\mathbf{x}_{0} \equiv \mathbf{x}\left(s_{0}\right) \quad \text { of } \quad \mathbf{G}(\mathbf{x})=\mathbf{0}
$$

is called a simple singular point if

$$
\mathrm{G}_{\mathbf{x}}^{0} \equiv \mathbf{G}_{\mathbf{x}}\left(\mathrm{x}_{0}\right) \quad \text { has rank } n-1
$$

In the parameter formulation, where

$$
\mathbf{G}_{\mathbf{x}}^{0}=\left(\mathbf{G}_{u}^{0} \mid \mathbf{G}_{\lambda}^{0}\right),
$$

we have that

$$
\mathbf{x}_{0}=\left(\mathbf{u}_{0}, \lambda_{0}\right) \quad \text { is a simple singular point }
$$

if and only if
(i) $\quad \operatorname{dim} \mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=1, \quad \mathbf{G}_{\lambda}^{0} \in \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)$,
or
(ii) $\quad \operatorname{dim} \mathcal{N}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)=2, \quad \mathbf{G}_{\lambda}^{0} \notin \mathcal{R}\left(\mathbf{G}_{\mathbf{u}}^{0}\right)$.


Figure 16: Solution curves of $u(\lambda-u)=0$, with simple singular point.

An example of case (ii) is

$$
\mathbf{G}(\mathbf{u}, \lambda)=\binom{\lambda-u_{1}^{2}-u_{2}^{2}}{u_{1} u_{2}}, \quad \text { at } \quad \lambda=0, \quad u_{1}=u_{2}=0
$$

Here

$$
\mathbf{G}_{\mathbf{u}}^{0}=\left(\begin{array}{ll}
0 & 0 \\
0 & 0
\end{array}\right), \quad \text { and } \quad \mathrm{G}_{\lambda}^{0}=\binom{1}{0}
$$



Figure 17: Solution curves of $\mathbf{G}(\mathbf{u}, \lambda)=\mathbf{0}$, with simple singular point.

Suppose we have a solution family $\mathbf{x}(s)$ of

$$
\mathbf{G}(\mathbf{x})=0,
$$

where $s$ is some parametrization.

Let

$$
\mathbf{x}_{0} \equiv\left(\mathbf{u}_{0}, \lambda_{0}\right)
$$

be a simple singular point.
Thus, by definition of simple singular point,

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Span}\left\{\boldsymbol{\phi}_{1}, \boldsymbol{\phi}_{2}\right\}, \quad \mathcal{N}\left(\mathbf{G}_{\mathrm{x}}^{0^{*}}\right)=\operatorname{Span}\{\boldsymbol{\psi}\} .
$$

We also have

$$
\begin{array}{ll}
\mathbf{G}(\mathbf{x}(s))=\mathbf{0}, & \mathbf{G}^{0}=\mathbf{G}\left(\mathbf{x}_{0}\right)=\mathbf{0} \\
\mathbf{G}_{\mathbf{x}}(\mathbf{x}(s)) \dot{\mathbf{x}}(s)=\mathbf{0}, & \mathbf{G}_{\mathbf{x}}^{0} \dot{\mathbf{x}}_{0}=\mathbf{0}, \\
\mathbf{G}_{\mathbf{x x}}(\mathbf{x}(s)) \dot{\mathbf{x}}(s) \dot{\mathbf{x}}(s)+\mathbf{G}_{\mathbf{x}}(\mathbf{x}(s)) \ddot{\mathbf{x}}(s)=\mathbf{0}, & \mathbf{G}_{\mathbf{x x}}^{0} \dot{\mathbf{x}}_{0} \dot{\mathbf{x}}_{0}+\mathbf{G}_{\mathbf{x}}^{0} \ddot{\mathbf{x}}_{0}=\mathbf{0} .
\end{array}
$$

Thus $\dot{\mathbf{x}}_{0}=\alpha \phi_{1}+\beta \phi_{2}$, for some $\alpha, \beta \in \mathrm{R}$, and

$$
\boldsymbol{\psi}^{*} \mathrm{G}_{\mathbf{x x}}^{0}\left(\alpha \boldsymbol{\phi}_{1}+\beta \boldsymbol{\phi}_{2}\right)\left(\alpha \boldsymbol{\phi}_{1}+\beta \boldsymbol{\phi}_{2}\right)+\underbrace{\boldsymbol{\psi}^{*} \mathrm{G}_{\mathbf{x}}^{0}}_{=0} \ddot{\mathbf{x}}_{0}=0
$$

$$
\underbrace{\left(\boldsymbol{\psi}^{*} \mathbf{G}_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1} \boldsymbol{\phi}_{1}\right)}_{c_{11}} \alpha^{2}+2 \underbrace{\left(\boldsymbol{\psi}^{*} \mathrm{G}_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1} \boldsymbol{\phi}_{2}\right)}_{c_{12}} \alpha \beta+\underbrace{\left(\boldsymbol{\psi}^{*} \mathbf{G}_{\mathbf{x x}}^{0} \boldsymbol{\phi}_{2} \boldsymbol{\phi}_{2}\right)}_{c_{22}} \beta^{2}=0 .
$$

This is the Algebraic Bifurcation equation (ABE).

We want solution pairs

$$
(\alpha, \beta)
$$

of the ABE , with not both $\alpha$ and $\beta$ equal to zero.

If the discriminant

$$
\Delta \equiv c_{12}^{2}-c_{11} c_{22}
$$

satisfies

$$
\Delta>0
$$

then the ABE has two real, distinct (i.e., linearly independent) solutions,

$$
\left(\alpha_{1}, \beta_{1}\right) \quad \text { and } \quad\left(\alpha_{2}, \beta_{2}\right),
$$

which are unique up to scaling.

In this case we have a bifurcation, (or branch point), i.e., two distinct branches pass through $\mathbf{x}_{0}$.

## Examples of Bifurcations

First we construct the Algebraic Bifurcation Equation (ABE) for our simple predator-prey model, in order to illustrate the necessary algebraic manipulations. Thereafter we present a more elaborate application to a nonlinear eigenvalue problem

## A Predator-Prey Model

In the 2-species predator-prey model

$$
\left\{\begin{array}{l}
u_{1}^{\prime}=3 u_{1}\left(1-u_{1}\right)-u_{1} u_{2}-\lambda\left(1-e^{-5 u_{1}}\right) \\
u_{2}^{\prime}=-u_{2}+3 u_{1} u_{2}
\end{array}\right.
$$

we have

$$
\mathbf{G}_{\mathbf{x}}=\left(\mathbf{G}_{u_{1}}\left|\mathbf{G}_{u_{2}}\right| \mathbf{G}_{\lambda}\right)=\left(\begin{array}{ccc}
3-6 u_{1}-u_{2}-5 \lambda e^{-5 u_{1}} & -u_{1} & -\left(1-e^{-5 u_{1}}\right) \\
3 u_{2} & -1+3 u_{1} & 0
\end{array}\right)
$$

and

$$
\mathbf{G}_{\mathbf{x x}}=\left(\begin{array}{lll}
\left(-6+25 \lambda e^{-5 u_{1}},-1,-5 e^{-5 u_{1}}\right) & (-1,0,0) & \left(-5 e^{-5 u_{1}}, 0,0\right) \\
(0,3,0) & (3,0,0) & (0,0,0)
\end{array}\right)
$$



Figure 18: Two branch points (Solutions 2 and 4) in AUTO demo pp2.

At Solution 2 in Figure 18 we have $u_{1}=u_{2}=0, \lambda=3 / 5$, so that

$$
\begin{gathered}
\mathbf{x}_{0}=(0,0,3 / 5), \\
\mathbf{G}_{\mathbf{x}}^{0}=\left(\begin{array}{rrr}
0 & 0 & 0 \\
0 & -1 & 0
\end{array}\right) \quad, \quad \mathbf{G}_{\mathrm{x}}^{0^{*}}=\left(\begin{array}{rr}
0 & 0 \\
0 & -1 \\
0 & 0
\end{array}\right), \\
\mathcal{N}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Span}\left\{\left(\begin{array}{l}
0 \\
0 \\
1
\end{array}\right),\left(\begin{array}{l}
1 \\
0 \\
0
\end{array}\right)\right\} \quad, \quad \mathcal{N}\left(\mathbf{G}_{\mathrm{x}}^{0^{*}}\right)=\operatorname{Span}\left\{\binom{1}{0}\right\},
\end{gathered}
$$

and

$$
\begin{gathered}
\mathrm{G}_{\mathrm{xx}}^{0}=\left(\begin{array}{lll}
(9,-1,-5) & (-1,0,0) & (-5,0,0) \\
(0,3,0) & (3,0,0) & (0,0,0)
\end{array}\right) . \\
\mathrm{G}_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1}=\left(\begin{array}{rrr}
-5 & 0 & 0 \\
0 & 0 & 0
\end{array}\right) \quad, \quad \mathrm{G}_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{2}=\left(\begin{array}{ccc}
9 & -1 & -5 \\
0 & 3 & 0
\end{array}\right) .
\end{gathered}
$$

Thus

$$
\begin{aligned}
\boldsymbol{\psi}^{*} \mathrm{G}_{\mathrm{xx}}^{0} \phi_{1} \boldsymbol{\phi}_{1} & =\boldsymbol{\psi}^{*}\left(\begin{array}{rrr}
-5 & 0 & 0 \\
0 & 0 & 0
\end{array}\right) \boldsymbol{\phi}_{1}=\boldsymbol{\psi}^{*}\binom{0}{0}=0 \\
\boldsymbol{\psi}^{*} \mathrm{G}_{\mathrm{xx}}^{0} \phi_{1} \phi_{2} & =\boldsymbol{\psi}^{*}\left(\begin{array}{rrr}
-5 & 0 & 0 \\
0 & 0 & 0
\end{array}\right) \boldsymbol{\phi}_{2}=\boldsymbol{\psi}^{*}\binom{-5}{0}=-5 \\
\boldsymbol{\psi}^{*} \mathrm{G}_{\mathrm{xx}}^{0} \phi_{2} \phi_{2} & =\boldsymbol{\psi}^{*}\left(\begin{array}{rrr}
9 & -1 & -5 \\
0 & 3 & 0
\end{array}\right) \boldsymbol{\phi}_{2}=\boldsymbol{\psi}^{*}\binom{9}{0}=9
\end{aligned}
$$

Therefore the ABE is

$$
-10 \alpha \beta+9 \beta^{2}=0
$$

which has two linearly independent solutions, namely,

$$
\binom{\alpha}{\beta}=\binom{1}{0}, \quad\binom{9}{10}
$$

Thus the (non-normalized) directions of the two bifurcation families at $\mathbf{x}_{0}$ are

$$
\dot{\mathbf{x}}_{0}=(1) \phi_{1}+(0) \phi_{2}=\left(\begin{array}{l}
0 \\
0 \\
1
\end{array}\right)=\left(\begin{array}{c}
\dot{u}_{1} \\
\dot{u}_{2} \\
\dot{\lambda}
\end{array}\right)
$$

and

$$
\dot{\mathbf{x}}_{0}=(9) \phi_{1}+(10) \phi_{2}=\left(\begin{array}{c}
10 \\
0 \\
9
\end{array}\right)=\left(\begin{array}{c}
\dot{u}_{1} \\
\dot{u}_{2} \\
\dot{\lambda}
\end{array}\right) \text {. }
$$

NOTE:

- The first direction is that of the zero solution family.
- The second direction is that of the bifurcating nonzero solution family.
- Since $\dot{\lambda} \neq 0$ for the second direction, this is a transcritical bifurcation.
- (The case where $\dot{\lambda}=0$ may correspond to a pitchfork bifurcation .)


## A Nonlinear Eigenvalue Problem

This example makes extensive use of the method of "Variation of Parameters" for solving linear differential equations. We first recall the use of this method.

## Variation of Parameters

If

$$
\mathbf{v}^{\prime}(t)=A(t) \mathbf{v}(t)+\mathbf{f}(t)
$$

then

$$
\mathbf{v}(t)=V(t)\left[\mathbf{v}(0)+\int_{0}^{t} V(s)^{-1} \mathbf{f}(s) d s\right]
$$

where $V(t)$ is the solution (matrix) of

$$
\begin{aligned}
V^{\prime}(t) & =A(t) V(t) \\
V(0) & =I
\end{aligned}
$$

Here $V(t)$ is called the fundamental solution matrix.

## EXAMPLE:

Apply Variation of Parameters to the equation

$$
v^{\prime \prime}+\lambda v=f
$$

rewritten as

$$
\left\{\begin{array}{l}
v_{1}^{\prime}=v_{2} \\
v_{2}^{\prime}=-\lambda v_{1}+f
\end{array}\right.
$$

or

$$
\binom{v_{1}^{\prime}}{v_{2}^{\prime}}=\left(\begin{array}{cc}
0 & 1 \\
-\lambda & 0
\end{array}\right)\binom{v_{1}}{v_{2}}+\binom{0}{f}
$$

Then

$$
V(t)=\left(\begin{array}{rl}
\cos \sqrt{\lambda} t & \sin \sqrt{\lambda} t / \sqrt{\lambda} \\
-\sqrt{\lambda} \sin \sqrt{\lambda} t & \cos \sqrt{\lambda} t
\end{array}\right)
$$

We find that

$$
\begin{gathered}
V^{-1}(t)=\left(\begin{array}{cc}
\cos \sqrt{\lambda} t & -\sin \sqrt{\lambda} t / \sqrt{\lambda} \\
\sqrt{\lambda} \sin \sqrt{\lambda} t & \cos \sqrt{\lambda} t
\end{array}\right), \\
V^{-1}(s)\binom{0}{f}=\binom{-\sin \sqrt{\lambda} s f(s) / \sqrt{\lambda}}{\cos \sqrt{\lambda} s f(s)},
\end{gathered}
$$

so that $\binom{v_{1}(t)}{v_{2}(t)}=$

$$
\left(\begin{array}{cc}
\cos \sqrt{\lambda} t & \sin \sqrt{\lambda} t / \sqrt{\lambda} \\
-\sqrt{\lambda} \sin \sqrt{\lambda} t & \cos \sqrt{\lambda} t
\end{array}\right)\left[\binom{v_{1}(0)}{v_{2}(0)}+\int_{0}^{t}\binom{-\sin \sqrt{\lambda} s f(s) / \sqrt{\lambda}}{\cos \sqrt{\lambda} s f(s)} d s\right]
$$

Hence

$$
\begin{aligned}
v_{1}(t) & =\cos (\sqrt{\lambda} t) v_{1}(0)+\frac{\sin (\sqrt{\lambda} t)}{\sqrt{\lambda}} v_{2}(0) \\
& -\cos \sqrt{\lambda} t \int_{0}^{t} \frac{\sin \sqrt{\lambda} s f(s)}{\sqrt{\lambda}} d s+\frac{\sin \sqrt{\lambda} t}{\sqrt{\lambda}} \int_{0}^{t} \cos \sqrt{\lambda} s f(s) d s
\end{aligned}
$$

For the specific initial value problem

$$
\left\{\begin{array}{l}
v^{\prime \prime}+\lambda v=f \\
v(0)=v^{\prime}(0)=0
\end{array}\right.
$$

we have

$$
v(t)=v_{1}(t)=\frac{\sin \sqrt{\lambda} t}{\sqrt{\lambda}} \int_{0}^{t} \cos \sqrt{\lambda} s f(s) d s-\frac{\cos \sqrt{\lambda} t}{\sqrt{\lambda}} \int_{0}^{t} \sin \sqrt{\lambda} s f(s) d s
$$

## Singular points

Consider the nonlinear boundary value problem

$$
\left\{\begin{array}{c}
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \\
u(0)=u(1)=0
\end{array}\right.
$$

which has $u(t) \equiv 0$ as solution for all $\lambda$.
Equivalently, we want the solution

$$
u=u(t ; p, \lambda),
$$

of the initial value problem

$$
\left\{\begin{array}{l}
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \\
u(0)=0, \quad u^{\prime}(0)=p
\end{array}\right.
$$

that satisfies

$$
G(p, \lambda) \equiv u(1 ; p, \lambda)=0 .
$$

$$
\begin{gathered}
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \quad, \quad u(0)=0 \quad, \quad u^{\prime}(0)=p \\
G(p, \lambda) \equiv u(1 ; p, \lambda)=0
\end{gathered}
$$

We have that

$$
G: \mathrm{R} \times \mathrm{R} \rightarrow \mathrm{R}
$$

with

$$
G(0, \lambda)=0, \quad \text { for all } \lambda
$$

Let

$$
\begin{gathered}
u_{p}(t ; p, \lambda)=\frac{d u}{d p}(t ; p, \lambda) \\
G_{p}(p, \lambda)=u_{p}(1 ; p, \lambda), \quad \text { etc. }
\end{gathered}
$$

$$
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \quad u(0)=0 \quad, \quad u^{\prime}(0)=p
$$

Then $u_{p}\left(=u_{p}^{0}\right)$ satisfies

$$
\left\{\begin{array}{l}
u_{p}^{\prime \prime}+\lambda(1+2 u) u_{p}=0 \\
u_{p}(0)=0, \quad u_{p}^{\prime}(0)=1
\end{array}\right.
$$

which, about $u \equiv 0$, gives

$$
\left\{\begin{array}{l}
u_{p}^{\prime \prime}+\lambda u_{p}=0 \\
u_{p}(0)=0, \quad u_{p}^{\prime}(0)=1
\end{array}\right.
$$

By the variation of parameters formula

$$
\begin{gathered}
u_{p}(t ; p, \lambda)=\frac{\sin \sqrt{\lambda} t}{\sqrt{\lambda}}, \quad \lambda \geq 0, \quad(\text { independent of } p) \\
G_{p}(0, \lambda)=u_{p}(1 ; p, \lambda)=\frac{\sin (\sqrt{\lambda})}{\sqrt{\lambda}}=0, \quad \text { if } \lambda=\lambda_{k} \equiv(k \pi)^{2}
\end{gathered}
$$

(We will see that the $\lambda_{k}$ are branch points.)

$$
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \quad u(0)=0 \quad, \quad u^{\prime}(0)=p
$$

Next, $u_{\lambda}$ satisfies

$$
\left\{\begin{array}{l}
u_{\lambda}^{\prime \prime}+u+u^{2}+\lambda(1+2 u) u_{\lambda}=0 \\
u_{\lambda}(0)=0, \quad u_{\lambda}^{\prime}(1)=0
\end{array}\right.
$$

which, about $u \equiv 0$, gives

$$
\left\{\begin{array}{l}
u_{\lambda}^{\prime \prime}+\lambda u_{\lambda}=0 \\
u_{\lambda}(0)=0, \quad u_{\lambda}^{\prime}(0)=0
\end{array}\right.
$$

from which,

$$
u_{\lambda}(t ; p, \lambda) \equiv 0, \quad G_{\lambda}(0, \lambda)=u_{\lambda}(1 ; 0, \lambda)=0
$$

which holds, in particular, at

$$
\lambda=\lambda_{k}=(k \pi)^{2}
$$

Thus, so far we know that

$$
G(0, \lambda)=0, \quad \text { for all } \lambda,
$$

and if

$$
\lambda=\lambda_{k} \equiv(k \pi)^{2},
$$

then, with

$$
\mathbf{x} \equiv(p, \lambda),
$$

we have

$$
\begin{gathered}
G_{\mathbf{x}}^{0} \equiv\left(G_{p}\left(0, \lambda_{k}\right) \mid G_{\lambda}\left(0, \lambda_{k}\right)\right)=(0 \mid 0), \\
\mathcal{N}\left(G_{\mathbf{x}}^{0}\right)=\operatorname{Span}\left\{\binom{1}{0},\binom{0}{1}\right\}, \\
\mathcal{N}\left(G_{\mathbf{x}}^{0^{*}}\right)=\operatorname{Span}\{(1)\},
\end{gathered}
$$

Thus the solutions

$$
p=0, \quad \lambda=\lambda_{k}=(k \pi)^{2},
$$

correspond to simple singular points.

## Construction of the ABE

The ABE is

$$
\left(\boldsymbol{\psi}^{*} G_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1} \boldsymbol{\phi}_{1}\right) \alpha^{2}+2\left(\boldsymbol{\psi}^{*} G_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1} \boldsymbol{\phi}_{2}\right) \alpha \beta+\left(\boldsymbol{\psi}^{*} G_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{2} \boldsymbol{\phi}_{2}\right) \beta^{2}=0
$$

where

$$
\begin{gathered}
\psi=1 \\
G_{\mathrm{xx}}^{0}=\left(\left(G_{p p}^{0} \mid G_{p \lambda}^{0}\right) \mid\left(G_{\lambda p}^{0} \mid G_{\lambda \lambda}^{0}\right)\right)
\end{gathered}
$$

with

$$
G_{p p}^{0}=G_{p p}\left(0, \lambda_{k}\right)=u_{p p}\left(1 ; 0, \lambda_{k}\right), \quad \text { etc. }
$$

If the ABE has 2 independent solutions then the bifurcation directions are

$$
\binom{\dot{p}}{\dot{\lambda}}=\alpha_{1}\binom{1}{0}+\beta_{1}\binom{0}{1}=\binom{\alpha_{1}}{\beta_{1}},
$$

and

$$
\binom{\dot{p}}{\dot{\lambda}}=\alpha_{2}\binom{1}{0}+\beta_{2}\binom{0}{1}=\binom{\alpha_{2}}{\beta_{2}} .
$$

Since

$$
p=0
$$

is a solution for all $\lambda$, one direction is

$$
\binom{\dot{p}}{\dot{\lambda}}=\binom{0}{1} .
$$

$$
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \quad u(0)=0 \quad, \quad u^{\prime}(0)=p
$$

Now $u_{p p}$ satisfies

$$
\left\{\begin{array}{l}
u_{p p}^{\prime \prime}+2 \lambda u_{p}^{2}+\lambda(1+2 u) u_{p p}=0 \\
u_{p p}(0)=u_{p p}^{\prime}(0)=0
\end{array}\right.
$$

which, about

$$
u \equiv 0, \quad \lambda=\lambda_{k}, \quad u_{p}(t ; p, \lambda)=\frac{\sin \sqrt{\lambda} t}{\sqrt{\lambda}}
$$

gives

$$
\left\{\begin{array}{l}
u_{p p}^{\prime \prime}+\lambda_{k} u_{p p}=-2 \lambda_{k} u_{p}^{2}=-2 \sin ^{2} \sqrt{\lambda} t \\
u_{p p}(0)=u_{p p}^{\prime}(0)=0
\end{array}\right.
$$

that is,

$$
\left\{\begin{array}{c}
u_{p p}^{\prime \prime}+(k \pi)^{2} u_{p p}=-2 \sin ^{2}(k \pi t) \\
u_{p p}(0)=u_{p p}^{\prime}(0)=0
\end{array}\right.
$$

By the variation of parameters formula

$$
\begin{aligned}
& u_{p p}(t ; p, \lambda)= \\
& \frac{\sin k \pi t}{k \pi} \int_{0}^{t} \cos k \pi s\left(-2 \sin ^{2} k \pi s\right) d s-\frac{\cos k \pi t}{k \pi} \int_{0}^{t} \sin k \pi s\left(-2 \sin ^{2} k \pi s\right) d s \\
& \quad=-\frac{2 \sin k \pi t}{k \pi} \int_{0}^{t} \sin ^{2} k \pi s \cos k \pi s d s+\frac{2 \cos k \pi t}{k \pi} \int_{0}^{t} \sin ^{3} k \pi s d s,
\end{aligned}
$$

and hence

$$
G_{p p}\left(0, \lambda_{k}\right)=u_{p p}\left(1 ; 0, \lambda_{k}\right)=\frac{2}{3(k \pi)^{2}}\left[1-(-1)^{k}\right]=\left\{\begin{array}{cc}
0, & k \text { even } \\
\frac{4}{3(k \pi)^{2}}, & k \text { odd }
\end{array}\right.
$$

$$
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \quad u(0)=0 \quad, \quad u^{\prime}(0)=p
$$

Next, $u_{p \lambda}$ satisfies

$$
\left\{\begin{array}{l}
u_{p \lambda}^{\prime \prime}+(1+2 u) u_{p}+2 \lambda u_{p} u_{\lambda}+\lambda(1+2 u) u_{p \lambda}=0 \\
u_{p \lambda}(0)=u_{p \lambda}^{\prime}(0)=0
\end{array}\right.
$$

which, about

$$
u \equiv 0, \quad \lambda_{k}=(k \pi)^{2}, \quad u_{p}(t ; p, \lambda)=\frac{\sin \sqrt{\lambda} t}{\sqrt{\lambda}}, \quad u_{\lambda}=0
$$

gives

$$
\left\{\begin{array}{l}
u_{p \lambda}^{\prime \prime}+k^{2} \pi^{2} u_{p \lambda}=-u_{p}=-\frac{\sin k \pi t}{k \pi}, \\
u_{p \lambda}(0)=u_{p \lambda}^{\prime}(0)=0
\end{array}\right.
$$

Using the variation of parameters formula

$$
u_{p \lambda}(t ; p, \lambda)=
$$

$$
\begin{aligned}
& \frac{\sin k \pi t}{k \pi} \int_{0}^{t} \cos k \pi s\left(\frac{-\sin k \pi s}{k \pi}\right) d s-\frac{\cos k \pi t}{k \pi} \int_{0}^{t} \sin k \pi s\left(\frac{-\sin k \pi s}{k \pi}\right) d s \\
= & \frac{\sin k \pi t}{(k \pi)^{2}} \int_{0}^{t} \sin k \pi s \cos k \pi s d s+\frac{\cos k \pi t}{(k \pi)^{2}} \int_{0}^{t} \sin ^{2} k \pi s d s,
\end{aligned}
$$

and hence

$$
G_{p \lambda}\left(0, \lambda_{k}\right)=u_{p \lambda}\left(1 ; 0, \lambda_{k}\right)=\frac{1}{2(k \pi)^{2}}(-1)^{k}= \begin{cases}\frac{1}{2(k \pi)^{2}}, & k \text { even }, \\ \frac{-1}{2(k \pi)^{2}}, & k \text { odd } .\end{cases}
$$

$$
u^{\prime \prime}+\lambda\left(u+u^{2}\right)=0 \quad u(0)=0 \quad, \quad u^{\prime}(0)=p
$$

Finally $u_{\lambda \lambda}$ satisfies

$$
\left\{\begin{array}{l}
u_{\lambda \lambda}^{\prime \prime}+(1+2 u) u_{\lambda}+(1+2 u) u_{\lambda}+2 \lambda u_{\lambda}^{2}+\lambda(1+2 u) u_{\lambda \lambda}=0 \\
u_{\lambda \lambda}(0)=u_{\lambda \lambda}^{\prime}(0)=0
\end{array}\right.
$$

which, about

$$
u=0, \quad u_{\lambda}=0, \quad \lambda=\lambda_{k}=k^{2} \pi^{2}
$$

gives

$$
\left\{\begin{array}{l}
u_{\lambda \lambda}^{\prime \prime}+k^{2} \pi^{2} u_{\lambda \lambda}=0 \\
u_{\lambda \lambda}(0)=u_{\lambda \lambda}^{\prime}(0)=0
\end{array}\right.
$$

so that

$$
u_{\lambda \lambda}(t ; p, \lambda) \equiv 0, \quad G_{\lambda \lambda}\left(0, \lambda_{k}\right)=u_{\lambda \lambda}\left(1 ; 0, \lambda_{k}\right)=0
$$

Thus we have found that

$$
G_{\mathbf{x x}}^{0}=\left(\left(G_{p p}^{0} \mid G_{p \lambda}^{0}\right) \mid\left(G_{\lambda p}^{0} \mid G_{\lambda \lambda}^{0}\right)\right)= \begin{cases}\left(\left.\left(0 \left\lvert\, \frac{1}{2(k \pi)^{2}}\right.\right) \right\rvert\,\left(\left.\frac{1}{2(k \pi)^{2}} \right\rvert\, 0\right)\right), & , k \text { even } \\ \left(\left.\left(\frac{4}{3(k \pi)^{2}} \left\lvert\, \frac{-1}{2(k \pi)^{2}}\right.\right) \right\rvert\,\left(\left.\frac{-1}{2(k \pi)^{2}} \right\rvert\, 0\right)\right), k \text { odd }\end{cases}
$$

The coefficients of the ABE are

$$
\begin{gathered}
\boldsymbol{\psi}^{*} G_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1} \boldsymbol{\phi}_{1}=\left\{\begin{array}{cc}
0, k \text { even }, \\
\frac{4}{3(k \pi)^{2}}, k \text { odd, }
\end{array} \quad \boldsymbol{\psi}^{*} G_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1} \boldsymbol{\phi}_{2}= \begin{cases}\frac{1}{2(k \pi)^{2}}, & k \text { even } \\
\frac{-1}{2(k \pi)^{2}}, & k \text { odd }\end{cases} \right. \\
\boldsymbol{\psi}^{*} G_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{2} \boldsymbol{\phi}_{2}= \begin{cases}0, & k \text { even } \\
0, & k \text { odd }\end{cases}
\end{gathered}
$$

Thus the ABE is

$$
\begin{cases}\alpha \beta=0, & k \text { even, } \\ \frac{\text { Roots }:(\alpha, \beta)=(1,0),(0,1)}{\frac{4}{3} \alpha^{2}-\frac{1}{2} \alpha \beta=0,} k \text { odd, } \quad \text { Roots }:(\alpha, \beta)=(0,1),(3,8)\end{cases}
$$

The directions of the bifurcating families are:

$$
\dot{\mathbf{x}}=\binom{\dot{p}}{\dot{\lambda}}=\alpha \phi_{1}+\beta \phi_{2}=\alpha\binom{1}{0}+\beta\binom{0}{1}
$$

where

$$
\binom{\dot{p}}{\dot{\lambda}}=\left\{\begin{array}{ll}
\binom{0}{1}, & \binom{1}{0},
\end{array} \quad k \text { even, },\left(\text { "pitch-fork bifurcation") }, ~\binom{0}{1},\binom{3}{8}, \quad k \text { odd, } \quad \text { ("transcritical bifurcation") } .\right.\right.
$$



Figure 19: The bifurcating families of the nonlinear eigenvalue problem.

## EXERCISE.

- Check the above calculations (!).
- Use the AUTO demo nev to compute some bifurcating families.
- Do the numerical results support the analytical results?
- Also carry out the above analysis and AUTO computations for

$$
\left\{\begin{array}{c}
u^{\prime \prime}+\lambda\left(u+u^{3}\right)=0 \\
u(0)=u(1)=0
\end{array}\right.
$$

## Branch Switching

- Along a solution family we may find branch points.
- We give two methods for switching branches.
- We also give a method to detect branch points.


## Computing the bifurcation direction

Let

$$
\mathrm{G}: \mathrm{R}^{n+1} \rightarrow \mathrm{R}^{n}
$$

Suppose that we have a solution family $\mathbf{x}(s)$ of

$$
\mathbf{G}(\mathbf{x})=\mathbf{0}
$$

and that

$$
\mathbf{x}_{0} \equiv \mathbf{x}\left(s_{0}\right)
$$

is a simple singular point, i.e., the $n$ by $n+1$ matrix

$$
\mathbf{G}_{\mathbf{x}}^{0} \equiv \mathbf{G}_{\mathbf{x}}\left(\mathbf{x}_{0}\right) \text { has rank } n-1
$$

with

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Span}\left\{\boldsymbol{\phi}_{1}, \boldsymbol{\phi}_{2}\right\}, \quad \mathcal{N}\left(\mathbf{G}_{\mathbf{x}}^{0^{*}}\right)=\operatorname{Span}\{\boldsymbol{\psi}\}
$$

## NOTATION.

- $\dot{\mathbf{x}}_{0}=\alpha_{1} \phi_{1}+\beta_{1} \phi_{2}$ denotes the direction of the "given" family.
- $\mathbf{x}_{0}^{\prime}=\alpha_{2} \phi_{1}+\beta_{2} \phi_{2}$ denotes the direction of the bifurcating family.

The two coefficient vectors

$$
\left(\alpha_{1}, \beta_{2}\right) \quad \text { and } \quad\left(\alpha_{2}, \beta_{2}\right)
$$

correspond to two linearly independent solutions of the ABE

$$
c_{11} \alpha^{2}+2 c_{12} \alpha \beta+c_{22} \beta^{2}=0
$$

Assume that the discriminant is positive:

$$
\Delta \equiv c_{12}^{2}-c_{11} c_{22}>0
$$

Since along the "given" family we have

$$
\mathrm{G}_{\mathrm{x}}^{0} \dot{\mathbf{x}}_{0}=\mathbf{0}
$$

we can take

$$
\phi_{1}=\dot{\mathbf{x}}_{0}, \quad\left(\dot{\mathbf{x}}_{0}=\alpha_{1} \phi_{1}+\beta_{1} \phi_{2}\right)
$$

Thus

$$
\left(\alpha_{1}, \beta_{1}\right)=(1,0)
$$

is a solution of the ABE

$$
c_{11} \alpha^{2}+2 c_{12} \alpha \beta+c_{22} \beta^{2}=0
$$

Thus

$$
c_{11}=0, \quad \text { and } \quad\left(\text { since } c_{12}^{2}-c_{11} c_{22}>0\right) \quad c_{12} \neq 0
$$

The second solution then satisfies

$$
2 c_{12} \alpha+c_{22} \beta=0
$$

from which,

$$
\left(\alpha_{2}, \beta_{2}\right)==\left(c_{22},-2 c_{12}\right), \quad \text { (unique, up to scaling) }
$$

To evaluate $c_{12}$ and $c_{22}$, we need the null vectors $\boldsymbol{\phi}_{1}$ and $\boldsymbol{\phi}_{2}$ of $\mathbf{G}_{\mathbf{x}}^{0}$.
$\phi_{1}$ : We already have chosen $\phi_{1}=\dot{\mathbf{x}}_{0}$.
$\phi_{2}$ : Choose $\phi_{2} \perp \phi_{1}$. Then $\phi_{2}$ is a null vector of

$$
\mathbf{F}_{\mathbf{x}}^{0}=\binom{\mathbf{G}_{\mathbf{x}}^{0}}{\dot{\mathbf{x}}_{0}^{*}} \quad \text { i.e., } \quad F_{\mathbf{x}}^{0} \phi_{2}=\binom{\mathbf{G}_{\mathbf{x}}^{0}}{\dot{\mathbf{x}}_{0}^{*}} \phi_{2}=\mathbf{0} .
$$

Note that $\mathbf{F}_{\mathbf{x}}^{0}$ is the Jacobian of the pseudo-arclength system at $\mathbf{x}_{0}$ !

The null space of $\mathbf{F}_{x}^{0}$ is indeed one-dimensional. (Check!)
$\boldsymbol{\psi}$ : is the left null vector: $\left(\mathrm{G}_{\mathrm{x}}^{0}\right)^{*} \boldsymbol{\psi}=\mathbf{0}$, so that also

$$
\left(\mathbf{F}_{\mathbf{x}}^{0}\right)^{*}\binom{\boldsymbol{\psi}}{0}=\left(\left(\mathbf{G}_{\mathbf{x}}^{0}\right)^{*} \mid \dot{\mathbf{x}}_{0}\right)\binom{\boldsymbol{\psi}}{0}=\mathbf{0}
$$

i.e., $\boldsymbol{\psi}$ is also the left null vector of $\mathbf{F}_{\mathbf{x}}^{0}$.

NOTE:

- Left and right null vectors of a matrix can be computed at little cost, once the matrix has been $L U$ decomposed.
- After determining the coefficients $\alpha_{2}$ and $\beta_{2}$, scale the direction vector

$$
\mathbf{x}_{0}^{\prime} \equiv \alpha_{2} \phi_{1}+\beta_{2} \phi_{2},
$$

of the bifurcating family so that

$$
\left\|\mathrm{x}_{0}^{\prime}\right\|=1
$$

## Switching branches

The first solution $\mathbf{x}_{1}$ on the bifurcating family can be computed from :

$$
\begin{gather*}
\mathbf{G}\left(\mathbf{x}_{1}\right)=\mathbf{0}, \\
\left(\mathbf{x}_{1}-\mathbf{x}_{0}\right)^{*} \mathbf{x}_{0}^{\prime}-\Delta s=0, \tag{3}
\end{gather*}
$$

where

$$
\mathrm{x}_{0}^{\prime} \text { is the direction of the bifurcating branch. }
$$

As initial approximation in Newton's method take

$$
\mathbf{x}_{1}^{(0)}=\mathbf{x}_{0}+\Delta s \mathbf{x}_{0}^{\prime} .
$$

For a graphical interpretation see Figure 20.
NOTE: Computing $\mathrm{x}_{0}^{\prime}$ requires evaluation of $\mathrm{G}_{\mathrm{xx}}^{0}$.


Figure 20: Switching branches using the correct bifurcation direction.

## Simplified branch switching

Instead of Eqn. (3) for the first solution on the bifurcating branch, use :

$$
\begin{gathered}
\mathbf{G}\left(\mathbf{x}_{1}\right)=\mathbf{0} \\
\left(\mathbf{x}_{1}-\mathbf{x}_{0}\right)^{*} \phi_{2}-\Delta s=0
\end{gathered}
$$

where $\phi_{2}$ is the second null vector of $\mathbf{G}_{\mathbf{x}}^{0}$, with, as before,

$$
\phi_{2} \perp \phi_{1}, \quad\left(\phi_{1}=\dot{\mathbf{x}}_{0}\right)
$$

i.e.,

$$
\binom{\mathbf{G}_{\mathrm{x}}^{0}}{\dot{\mathbf{x}}_{0}^{*}} \phi_{2}=0, \quad\left\|\boldsymbol{\phi}_{2}\right\|=1
$$

As initial approximation, now use

$$
\mathbf{x}_{1}^{(0)}=\mathbf{x}_{0}+\Delta s \phi_{2}
$$

For a graphical interpretation see Figure 21.


Figure 21: Switching branches using the orthogonal direction.

NOTE:

- The simplified branch switching method may fail in some situations.
- The advantage is that it does not need second derivatives .
- The orthogonal direction $\phi_{2}$ can be computed at little cost.
- In fact, $\phi_{2}$ is the null vector of the pseudo-arclength Jacobian

$$
\binom{\mathrm{G}_{\mathrm{x}}^{0}}{\dot{\mathbf{x}}_{0}^{*}}
$$

at the branch point.

## Detection of Branch Points

Let

$$
\mathrm{G}: \mathrm{R}^{n+1} \rightarrow \mathrm{R}^{n}
$$

Recall that a solution

$$
\mathbf{x}_{0} \equiv \mathbf{x}\left(s_{0}\right) \quad \text { of } \quad \mathbf{G}(\mathbf{x})=\mathbf{0}
$$

is a simple singular point if

$$
\mathrm{G}_{\mathrm{x}}^{0} \equiv \mathrm{G}_{\mathbf{x}}\left(\mathrm{x}_{0}\right) \quad \text { has rank } n-1
$$

Suppose that we have a solution family $\mathbf{x}(s)$ of

$$
\begin{aligned}
& \mathbf{G}(\mathbf{x})=\mathbf{0}, \\
& \mathbf{x}_{0}=\mathbf{x}(0),
\end{aligned}
$$

and that
is a simple singular point.
Let $\dot{\mathbf{x}}_{0}$ be the unit tangent to $\mathbf{x}(s)$ at $\mathbf{x}_{0}$.
Assume that $\mathbf{x}(s)$ is parametrized by its projection onto $\dot{\mathbf{x}}_{0}$. (See Figure 22.)


Figure 22: Parametrization of a solution family near a branch point.

Consider the pseudo-arclength system

$$
\begin{equation*}
\mathbf{F}(\mathbf{x} ; s) \equiv\binom{\mathbf{G}(\mathbf{x})}{\left(\mathbf{x}-\mathbf{x}_{0}\right)^{*} \dot{\mathbf{x}}_{0}-s} \tag{4}
\end{equation*}
$$

Then

$$
\mathbf{F}_{\mathbf{x}}(\mathbf{x} ; s)=\mathbf{F}_{\mathbf{x}}(\mathbf{x})=\binom{\mathbf{G}_{\mathbf{x}}(\mathbf{x})}{\dot{\mathbf{x}}_{0}^{*}}
$$

and

$$
\mathbf{F}_{\mathbf{x}}^{0} \equiv \mathbf{F}_{\mathbf{x}}\left(\mathbf{x}_{0}\right)=\binom{\mathbf{G}_{\mathrm{x}}^{0}}{\dot{\mathbf{x}}_{0}^{*}}
$$

NOTE: $\quad \mathbf{F}_{\mathbf{x}}$ does not explicitly depend on $s$.

Take

$$
\phi_{1}=\dot{\mathbf{x}}_{0}
$$

as the first null vector of $\mathrm{G}_{\mathrm{x}}^{0}$.
Thus

$$
\mathbf{F}_{\mathbf{x}}^{0}=\binom{\mathbf{G}_{\mathbf{x}}^{0}}{\boldsymbol{\phi}_{1}^{*}}
$$

Choose the second null vector $\phi_{2}$ of $\mathrm{G}_{\mathrm{x}}^{0}$ such that

$$
\phi_{2}^{*} \phi_{1}=0
$$

Then

$$
\mathbf{F}_{\mathrm{x}}^{0} \boldsymbol{\phi}_{2}=\binom{\mathbf{G}_{\mathrm{x}}^{0}}{\boldsymbol{\phi}_{1}^{*}} \boldsymbol{\phi}_{2}=0
$$

so that $\phi_{2}$ is also a null vector of $\mathbf{F}_{\mathbf{x}}^{0}$, while $\boldsymbol{\phi}_{1}$ is not.

In fact, $\mathbf{F}_{\mathbf{x}}^{0}$ has a one-dimensional nullspace .

The null vector of

$$
\mathbf{F}_{\mathbf{x}}^{0 *}=\binom{\mathbf{G}_{\mathbf{x}}^{0}}{\boldsymbol{\phi}_{1}^{*}}^{*}=\left(\left(\mathbf{G}_{\mathbf{x}}^{0}\right)^{*} \mid \boldsymbol{\phi}_{1}\right)
$$

is given by

$$
\Psi \equiv\binom{\psi}{0}
$$

where $\boldsymbol{\psi}$ is the null vector of $\left(\mathbf{G}_{\mathrm{x}}^{0}\right)^{*}$.

NOTE: Since

$$
\mathrm{G}_{\mathbf{x}}^{0} \text { has } n \text { rows and } n+1 \text { columns, }
$$

and

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{x}}^{0}\right)=\operatorname{Span}\left\{\phi_{1}, \phi_{2}\right\} \quad \text { is assumed two-dimensional }
$$

it follows that

$$
\left(\mathbf{G}_{\mathbf{x}}^{0}\right)^{*} \text { has } n+1 \text { rows and } n \text { columns }
$$

and

$$
\mathcal{N}\left(\mathbf{G}_{\mathbf{x}}^{0^{*}}\right)=\operatorname{Span}\{\boldsymbol{\psi}\} \quad \text { is one-dimensional }
$$

THEOREM. Let

$$
\mathbf{x}_{0}=\mathbf{x}(0),
$$

be a simple singular point on a smooth solution family $\mathbf{x}(s)$ of

$$
\mathbf{G}(\mathbf{x})=\mathbf{0}
$$

Let $\mathbf{F}(\mathbf{x} ; s)$ be as above, i.e., $\quad \mathbf{F}(\mathbf{x} ; s) \equiv\binom{\mathbf{G}(\mathbf{x})}{\left(\mathbf{x}-\mathbf{x}_{0}\right)^{*} \dot{\mathbf{x}}_{0}-s}$.
Assume that

- the discriminant $\Delta$ of the ABE is positive,
- 0 is an algebraically simple eigenvalue of

$$
\mathbf{F}_{\mathbf{x}}^{0} \equiv\binom{\mathbf{G}_{\mathbf{x}}^{0}}{\dot{\mathbf{x}}_{0}^{*}} .
$$

Then

$$
\operatorname{det} \mathbf{F}_{\mathbf{x}}(\mathbf{x}(s))=\operatorname{det}\binom{\mathbf{G}_{\mathbf{x}}(\mathbf{x}(s))}{\dot{\mathbf{x}}_{0}^{*}}
$$

changes sign at $\mathbf{x}_{0}$.

## PROOF.

Consider the parametrized eigenvalue problem

$$
\mathbf{F}_{\mathbf{x}}(\mathbf{x}(s)) \boldsymbol{\phi}(s)=\kappa(s) \boldsymbol{\phi}(s),
$$

where $\kappa(s)$ and $\boldsymbol{\phi}(s)$ are smooth near $s=0$, with

$$
\kappa(0)=0 \quad \text { and } \quad \phi(0)=\phi_{2},
$$

i.e., the eigen pair

$$
(\kappa(s), \boldsymbol{\phi}(s)),
$$

is the continuation of $\left(0, \phi_{2}\right)$.
(This can be done because 0 is an algebraically simple eigenvalue.)

Differentiating

$$
\mathbf{F}_{\mathbf{x}}(\mathbf{x}(s)) \boldsymbol{\phi}(s)=\kappa(s) \boldsymbol{\phi}(s)
$$

gives

$$
\mathbf{F}_{\mathbf{x} \mathbf{x}}(\mathbf{x}(s)) \dot{\mathbf{x}}(s) \boldsymbol{\phi}(s)+\mathbf{F}_{\mathbf{x}}(\mathbf{x}(s)) \dot{\boldsymbol{\phi}}(s)=\dot{\kappa}(s) \boldsymbol{\phi}(s)+\kappa(s) \dot{\boldsymbol{\phi}}(s)
$$

Evaluating at $s=0$, using

$$
\kappa(0)=0
$$

and

$$
\dot{\mathbf{x}}_{0}=\phi_{1} \quad \text { and } \quad \phi(0)=\phi_{2}
$$

gives

$$
\mathbf{F}_{\mathrm{xx}}^{0} \phi_{1} \phi_{2}+\mathbf{F}_{\mathrm{x}}^{0} \dot{\phi}(0)=\dot{\kappa}_{0} \phi_{2}
$$

$$
\mathbf{F}_{\mathrm{xx}}^{0} \phi_{1} \phi_{2}+\mathbf{F}_{\mathrm{x}}^{0} \dot{\phi}(0)=\dot{\kappa}_{0} \phi_{2} .
$$

Multiplying this on the left by $\boldsymbol{\Psi}^{*}$ we find

$$
\dot{\kappa}_{0}=\frac{\boldsymbol{\Psi}^{*} \mathbf{F}_{\mathbf{x x}}^{0} \boldsymbol{\phi}_{1} \phi_{2}}{\boldsymbol{\Psi}^{*} \phi_{2}}=\frac{\left(\boldsymbol{\psi}^{*}, 0\right)\binom{\mathrm{G}_{\mathrm{xx}}^{0}}{0} \phi_{1} \phi_{2}}{\boldsymbol{\Psi}^{*} \boldsymbol{\phi}_{2}}=\frac{\boldsymbol{\psi}^{*} \mathrm{G}_{\mathrm{xx}}^{0} \phi_{1} \phi_{2}}{\boldsymbol{\Psi}^{*} \phi_{2}}
$$

The left and right null vectors ( $\mathbf{\Psi}$ and $\boldsymbol{\phi}_{2}$ ) of $\mathbf{F}_{\mathbf{x}}^{0} \quad$ cannot be orthogonal (because the eigenvalue 0 is assumed to be algebraically simple.

Thus

$$
\Psi^{*} \phi_{2} \neq 0
$$

Note that

$$
\boldsymbol{\psi}^{*} \mathrm{G}_{\mathrm{xx}}^{0} \boldsymbol{\phi}_{1} \boldsymbol{\phi}_{2}=c_{12}
$$

is a coefficient of the ABE .

By assumption, the discriminant satisfies

$$
\Delta \neq 0
$$

As before this implies that

$$
c_{12} \neq 0,
$$

and hence

$$
\dot{\kappa}_{0} \neq 0
$$

NOTE:

- The Theorem implies that $\operatorname{det}\binom{\mathbf{G}_{\mathbf{x}}(\mathbf{x}(s))}{\dot{\mathbf{x}}_{0}^{*}}$ changes sign.
- Note that $\dot{\mathbf{x}}_{0}$ is kept fixed.
- We haven't proved that $\operatorname{det} \mathbf{F}_{\mathbf{x}}(\mathbf{x}(s))=\operatorname{det}\binom{\mathbf{G}_{\mathbf{x}}(\mathbf{x}(s))}{\dot{\mathbf{x}}(s)^{*}}$ changes sign.
- The latter follows from a similar, but more elaborate argument.
- Detection of simple singular points is based upon this fact.
- During continuation we monitor the determinant of the matrix $\mathbf{F}_{\mathbf{x}}$.
- If a sign change is detected then an iterative method can be used to accurately locate the singular point.
- For large systems a scaled determinant avoids overflow.

The following theorem states that there must be a bifurcation at $x_{0}$. (This result can be proven by degree theory .)

## THEOREM.

Let $\mathbf{x}(s)$ be a smooth solution family of

$$
\mathbf{F}(\mathbf{x} ; s)=\mathbf{0}
$$

where

$$
\mathbf{F}: \mathrm{R}^{n+1} \times \mathrm{R} \quad \rightarrow \quad \mathrm{R}^{n+1} \quad \text { is } \quad C^{1}
$$

and assume that

$$
\operatorname{det} \mathbf{F}_{\mathbf{x}}(\mathbf{x}(s) ; s) \text { changes sign at } s=0 .
$$

Then $\mathbf{x}(0)$ is a bifurcation point, i.e., every open neigborhood of $\mathbf{x}_{0}$ contains a solution of $\mathbf{F}(\mathbf{x} ; s)=\mathbf{0}$ that does not lie on $\mathbf{x}(s)$.

## Boundary Value Problems

## Boundary Value Problems.

Consider the first order system of ordinary differential equations

$$
\mathbf{u}^{\prime}(t)-\mathbf{f}(\mathbf{u}(t), \mu, \lambda)=\mathbf{0}, \quad t \in[0,1]
$$

where

$$
\mathbf{u}(\cdot), \mathbf{f}(\cdot) \in \mathrm{R}^{n}, \quad \lambda \in \mathrm{R}, \quad \mu \in \mathrm{R}^{n_{\mu}}
$$

subject to boundary conditions

$$
\mathbf{b}(\mathbf{u}(0), \mathbf{u}(1), \mu, \lambda)=\mathbf{0}, \quad \mathbf{b}(\cdot) \in \mathrm{R}^{n_{b}}
$$

and integral constraints

$$
\int_{0}^{1} \mathbf{q}(\mathbf{u}(s), \mu, \lambda) d s=\mathbf{0}, \quad \mathbf{q}(\cdot) \in \mathrm{R}^{n_{q}}
$$

This boundary value problem (BVP) is of the form

$$
\mathbf{F}(\mathbf{X})=0
$$

where

$$
\mathbf{X}=(\mathbf{u}, \mu, \lambda),
$$

to which we add the pseudo-arclength equation

$$
<\mathbf{X}-\mathbf{X}_{0}, \dot{\mathbf{X}}_{0}>-\Delta s=0
$$

where $\mathbf{X}_{0}$ represents the preceding solution on the branch.
In detail, the pseudo-arclength equation is

$$
\begin{aligned}
\int_{0}^{1}\left(\mathbf{u}(t)-\mathbf{u}_{0}(t)\right)^{*} \dot{\mathbf{u}}_{0}(t) d t+ & \left(\mu-\mu_{0}\right) \dot{\mu}_{0} \\
& +\left(\lambda-\lambda_{0}\right) \dot{\lambda}_{0}-\Delta s=0 .
\end{aligned}
$$

- We want to solve BVP for $\mathbf{u}(\cdot)$ and $\mu$.
- We can think of $\lambda$ as the continuation parameter .
- (In pseudo-arclength continuation, we don't distinguish $\mu$ and $\lambda$.)
- In order for problem to be formally well-posed we must have

$$
n_{\mu}=n_{b}+n_{q}-n \geq 0
$$

- A simple case is

$$
n_{q}=0, \quad n_{b}=n, \quad \text { for which } \quad n_{\mu}=0
$$

## Discretization

Here we discuss the method of "orthogonal collocation with piecewise polynomials", for solving boundary value problems. This method is very accurate, and allows adaptive mesh-selection.

## Orthogonal Collocation

Introduce a mesh

$$
\left\{0=t_{0}<t_{1}<\cdots<t_{N}=1\right\}
$$

where

$$
h_{j} \equiv t_{j}-t_{j-1}, \quad(1 \leq j \leq N)
$$

Define the space of (vector) piecewise polynomials $\mathbf{P}_{h}^{m}$ as

$$
\mathbf{P}_{h}^{m} \equiv\left\{\mathbf{p}_{h} \in C[0,1]:\left.\mathbf{p}_{h}\right|_{\left[t_{j-1}, t_{j}\right]} \in \mathbf{P}^{m}\right\}
$$

where $\mathbf{P}^{m}$ is the space of (vector) polynomials of degree $\leq m$.

The collocation method consists of finding

$$
\mathbf{p}_{h} \in \mathbf{P}_{h}^{m}, \quad \mu \in \mathrm{R}^{n_{\mu}},
$$

such that the following collocation equations are satisfied:

$$
\mathbf{p}_{h}^{\prime}\left(z_{j, i}\right)=\mathbf{f}\left(\mathbf{p}_{h}\left(z_{j, i}\right), \mu, \lambda\right), \quad j=1, \cdots, N, \quad i=1, \cdots, m,
$$

and such that $\mathbf{p}_{h}$ satisfies the boundary and integral conditions.

The collocation points $z_{j, i}$ in each subinterval

$$
\left[t_{j-1}, t_{j}\right]
$$

are the (scaled) roots of the $m$ th-degree orthogonal polynomial (Gauss points).
See Figure 23 for a graphical interpretation.


Figure 23: The mesh $\left\{0=t_{0}<t_{1}<\cdots<t_{N}=1\right\}$. Collocation points and "extended-mesh points" are shown for the case $m=3$, in the $j$ th mesh interval. Also shown are two of the four local Lagrange basis polynomials.

Since each local polynomial is determined by

$$
(m+1) n,
$$

coefficients, the total number of degrees of freedom (considering $\lambda$ as fixed) is

$$
(m+1) n N+n_{\mu} .
$$

This is matched by the total number of equations :

$$
\begin{array}{ll}
\text { collocation : } & m n N, \\
\text { continuity : } & (N-1) n, \\
\text { constraints : } & n_{b}+n_{q}\left(=n+n_{\mu}\right) .
\end{array}
$$

Assume that the solution $\mathbf{u}(t)$ of the BVP is sufficiently smooth.

Then the order of accuracy of the orthogonal collocation method is $m$, i.e.,

$$
\left\|\mathbf{p}_{h}-\mathbf{u}\right\|_{\infty}=\mathcal{O}\left(h^{m}\right)
$$

At the main meshpoints $t_{j}$ we have superconvergence:

$$
\max _{j}\left|\mathbf{p}_{h}\left(t_{j}\right)-\mathbf{u}\left(t_{j}\right)\right|=\mathcal{O}\left(h^{2 m}\right)
$$

The scalar variables $\mu$ are also superconvergent.

## Implementation

For each subinterval $\left[t_{j-1}, t_{j}\right]$, introduce the Lagrange basis polynomials

$$
\left\{\ell_{j, i}(t)\right\}, \quad j=1, \cdots, N, \quad i=0,1, \cdots, m
$$

defined by

$$
\ell_{j, i}(t)=\prod_{k=0, k \neq i}^{m} \frac{t-t_{j-\frac{k}{m}}}{t_{j-\frac{i}{m}}-t_{j-\frac{k}{m}}},
$$

where

$$
t_{j-\frac{i}{m}} \equiv t_{j}-\frac{i}{m} h_{j}
$$

The local polynomials can then be written

$$
\mathbf{p}_{j}(t)=\sum_{i=0}^{m} \ell_{j, i}(t) \mathbf{u}_{j-\frac{i}{m}}
$$

With the above choice of basis

$$
\mathbf{u}_{j} \sim \mathbf{u}\left(t_{j}\right) \quad \text { and } \quad \mathbf{u}_{j-\frac{i}{m}} \sim \mathbf{u}\left(t_{j-\frac{i}{m}}\right)
$$

where $\mathbf{u}(t)$ is the solution of the continuous problem.

The collocation equations are

$$
\mathbf{p}_{j}^{\prime}\left(z_{j, i}\right)=\mathbf{f}\left(\mathbf{p}_{j}\left(z_{j, i}\right), \mu, \lambda\right), \quad i=1, \cdots, m, \quad j=1, \cdots, N
$$

The discrete boundary conditions are

$$
b_{i}\left(\mathbf{u}_{0}, \mathbf{u}_{N}, \mu, \lambda\right)=0, \quad i=1, \cdots, n_{b}
$$

The integral constraints can be discretized as

$$
\sum_{j=1}^{N} \sum_{i=0}^{m} \omega_{j, i} q_{k}\left(\mathbf{u}_{j-\frac{i}{m}}, \mu, \lambda\right)=0, \quad k=1, \cdots, n_{q}
$$

where the $\omega_{j, i}$ are the Lagrange quadrature weights.

The pseudo-arclength equation is

$$
\int_{0}^{1}\left(\mathbf{u}(t)-\mathbf{u}_{0}(t)\right)^{*} \dot{\mathbf{u}}_{0}(t) d t+\left(\mu-\mu_{0}\right)^{*} \dot{\mu}_{0}+\left(\lambda-\lambda_{0}\right) \dot{\lambda}_{0}-\Delta s=0
$$

where

$$
\left(\mathbf{u}_{0}, \mu_{0}, \lambda_{0}\right)
$$

is the previous solution on the solution branch, and

$$
\left(\dot{\mathbf{u}}_{0}, \dot{\mu}_{0}, \dot{\lambda}_{0}\right),
$$

is the normalized direction of the branch at the previous solution.

The discretized pseudo-arclength equation is

$$
\begin{aligned}
& \sum_{j=1}^{N} \sum_{i=0}^{m} \omega_{j, i}\left[\mathbf{u}_{j-\frac{i}{m}}-\left(\mathbf{u}_{0}\right)_{j-\frac{i}{m}}\right]^{*}\left(\dot{\mathbf{u}}_{0}\right)_{j-\frac{i}{m}} \\
&+\left(\mu-\mu_{0}\right)^{*} \dot{\mu}_{0}+\left(\lambda-\lambda_{0}\right) \dot{\lambda}_{0}-\Delta s=0
\end{aligned}
$$

## Numerical Linear Algebra

The complete discretization consists of

$$
m n N+n_{b}+n_{q}+1
$$

nonlinear equations, in the unknowns

$$
\left\{\mathbf{u}_{j-\frac{i}{m}}\right\} \in \mathrm{R}^{m n N+n}, \quad \mu \in \mathrm{R}^{n_{\mu}}, \quad \lambda \in \mathrm{R}
$$

These equations can be solved by a Newton-Chord iteration.

We illustrate the numerical linear algebra for the case

$$
\begin{gathered}
n=2 \text { ODEs } \quad, \quad N=4 \text { mesh intervals }, m=3 \text { collocation points } \\
n_{b}=2 \text { boundary conditions }, n_{q}=1 \text { integral constraint }
\end{gathered}
$$

and the pseudo-arclength equation.

- The operations are also done on the right hand side, which is not shown.
- Entries marked "०" have been eliminated by Gauss elimination.
- Entries marked "." denote fill-in due to pivoting.
- Most of the operations can be done in parallel.


Figure 24: The structure of the Jacobian


Figure 25: The system after condensation of parameters.


Figure 26: The preceding matrix, showing the decoupled $\star$ sub-system.


Figure 27: Stage 1 of the nested dissection to solve the decoupled $\star$ system.


Figure 28: Stage 2 of the nested dissection to solve the decoupled $\star$ system.


Figure 29: The preceding matrix showing the final decoupled + sub-system.


Figure 30: The Floquet Multipliers are the eigenvalues of $-B^{-1} A$.

## Accuracy Test

The Table shows the location of the fold in the Gelfand-Bratu problem for

- 4 Gauss collocation points per mesh interval
- $N$ mesh intervals

| $N$ | Fold location |
| ---: | :--- |
| 2 | 3.5137897550 |
| 4 | 3.5138308601 |
| 8 | 3.5138307211 |
| 16 | 3.5138307191 |
| 32 | 3.5138307191 |

## A Singularly-Perturbed BVP

$$
\epsilon u^{\prime \prime}(x)=u(x) u^{\prime}(x)\left(u(x)^{2}-1\right)+u(x) .
$$

with boundary conditions

$$
u(0)=\frac{3}{2} \quad, \quad u(1)=\gamma
$$

Computational formulation

$$
\begin{aligned}
u_{1}^{\prime} & =u_{2} \\
u_{2}^{\prime} & =\frac{\lambda}{\epsilon}\left(u_{1} u_{2}\left(u_{1}^{2}-1\right)+u_{1}\right)
\end{aligned}
$$

with boundary conditions

$$
u_{1}(0)=3 / 2 \quad, \quad u_{1}(1)=\gamma
$$

When $\lambda=0$ an exact solution is

$$
u 1(x)=\frac{3}{2}+\left(\gamma-\frac{3}{2}\right) x \quad, \quad u_{2}(x)=\gamma-\frac{3}{2}
$$

## COMPUTATIONAL STEPS:

- $\quad \lambda$ is a homotopy parameter to locate a starting solution.
- In the first run $\lambda$ varies from 0 to 1 .
- In the second run $\epsilon$ is decreased by continuation.
- In the third run $\epsilon=10^{-3}$, and the solution is continued in $\gamma$.
- This third run takes many continuation steps if $\epsilon$ is very small.


Figure 31: Bifurcation diagram of the singularly-perturbed BVP.


Figure 32: Some solutions along the solution family.

## Hopf Bifurcation and Periodic Solutions

We first introduce the concept of Hopf bifurcation for a "linear" problem. Then we state the Hopf Bifurcation Theorem (without proof), and we give examples of periodic solutions emanating from Hopf bifurcation points.

## A Linear Example

The linear problem :

$$
\left\{\begin{array}{l}
u_{1}^{\prime}=\lambda u_{1}-u_{2}  \tag{5}\\
u_{2}^{\prime}=u_{1}
\end{array}\right.
$$

which can also be written as

$$
\binom{u_{1}^{\prime}}{u_{2}^{\prime}}=\left(\begin{array}{rr}
\lambda & -1 \\
1 & 0
\end{array}\right)\binom{u_{1}}{u_{2}},
$$

is of the form

$$
\mathbf{u}^{\prime}(t)=A(\lambda) \mathbf{u}(t) \equiv \mathbf{G}(\mathbf{u}, \lambda),
$$

with stationary solutions,

$$
u_{1}=u_{2}=0, \quad \text { for all } \lambda .
$$

The eigenvalues $\mu$ of the Jacobian matrix

$$
\mathbf{G}_{\mathbf{u}}(\mathbf{u}, \lambda)=A(\lambda)=\left(\begin{array}{rr}
\lambda & -1 \\
1 & 0
\end{array}\right)=\mathbf{G}_{\mathbf{u}}(\mathbf{0}, \lambda)
$$

satisfy

$$
\operatorname{det}(A(\lambda)-\mu I)=\mu^{2}-\lambda \mu+1=0,
$$

from which

$$
\mu_{1,2}=\frac{\lambda \pm \sqrt{\lambda^{2}-4}}{2}
$$

Consider the initial value problem

$$
\mathbf{u}^{\prime}(t)=A(\lambda) \mathbf{u}(t) \equiv \mathbf{G}(\mathbf{u}, \lambda),
$$

with

$$
\mathbf{u}(0)=\binom{u_{1}(0)}{u_{2}(0)}=\binom{p_{1}}{p_{2}} \equiv \mathbf{p} .
$$

Then

$$
\begin{align*}
\mathbf{u}(t) & =e^{t A(\lambda)} \mathbf{p}=V(\lambda) e^{t \Lambda(\lambda)} V^{-1}(\lambda) \mathbf{p} \\
& =V(\lambda)\left(\begin{array}{cc}
e^{t \mu_{1}(\lambda)} & 0 \\
0 & e^{t \mu_{2}(\lambda)}
\end{array}\right) V^{-1}(\lambda) \mathbf{p} \tag{6}
\end{align*}
$$

where

$$
A(\lambda) V(\lambda)=V(\lambda) \Lambda(\lambda), \quad A(\lambda)=V(\lambda) \Lambda(\lambda) V^{-1}(\lambda)
$$

and

$$
\Lambda(\lambda)=\left(\begin{array}{cc}
\mu_{1}(\lambda) & 0 \\
0 & \mu_{2}(\lambda)
\end{array}\right), \quad V(\lambda)=\left(\begin{array}{ll}
v_{11}(\lambda) & v_{12}(\lambda) \\
v_{21}(\lambda) & v_{22}(\lambda)
\end{array}\right)
$$

Assume that

$$
-2<\lambda<2
$$

and recall that

$$
\mathbf{u}(t)=V(\lambda)\left(\begin{array}{cc}
e^{t \mu_{1}(\lambda)} & 0 \\
0 & e^{t \mu_{2}(\lambda)}
\end{array}\right) V^{-1}(\lambda) \mathbf{p}
$$

and

$$
\mu_{1,2}=\frac{\lambda \pm \sqrt{\lambda^{2}-4}}{2} .
$$

Thus we see that

$$
\mathbf{u}(t) \rightarrow \mathbf{0} \quad \text { if } \quad \lambda<0
$$

and

$$
\mathbf{u}(t) \rightarrow \infty \quad \text { if } \quad \lambda>0
$$

i.e., the zero solution is stable if $\lambda$ is negative, and unstable if $\lambda$ is positive.

However, if $\lambda=0$, then

$$
A_{0} \equiv A(0)=\left(\begin{array}{rr}
0 & -1 \\
1 & 0
\end{array}\right)
$$

and

$$
\begin{gathered}
\mu_{1}=i, \quad \mu_{2}=-i, \\
V_{0} \equiv V(0)=\left(\begin{array}{rr}
1 & -i \\
-i & 1
\end{array}\right), \\
V_{0}^{-1}=\frac{1}{2}\left(\begin{array}{cc}
1 & i \\
i & 1
\end{array}\right),
\end{gathered}
$$

so that

$$
\begin{gathered}
\mathbf{u}(t)=V_{0} e^{t \Lambda} V_{0}^{-1} \mathbf{p}=\frac{1}{2}\left(\begin{array}{rr}
1 & -i \\
-i & 1
\end{array}\right)\left(\begin{array}{cc}
e^{i t} & 0 \\
0 & e^{-i t}
\end{array}\right)\left(\begin{array}{ll}
1 & i \\
i & 1
\end{array}\right)\binom{p_{1}}{p_{2}} \\
=\frac{1}{2}\left(\begin{array}{cc}
e^{i t}+e^{-i t} & i\left(e^{i t}-e^{-i t}\right) \\
-i\left(e^{i t}-e^{-i t}\right) & e^{i t}+e^{-i t}
\end{array}\right)\binom{p_{1}}{p_{2}} .
\end{gathered}
$$

Thus, if $\lambda=0$, then

$$
\binom{u_{1}(t)}{u_{2}(t)}=\left(\begin{array}{rr}
\cos (t) & -\sin (t) \\
\sin (t) & \cos (t)
\end{array}\right)\binom{p_{1}}{p_{2}}=\binom{p_{1} \cos (t)-p_{2} \sin (t)}{p_{1} \sin (t)+p_{2} \cos (t)},
$$

and we see that

- This solution is periodic, with period $2 \pi$, for any $p_{1}, p_{2}$.
- $u_{1}(t)^{2}+u_{2}(t)^{2}=p_{1}^{2}+p_{2}^{2}, \quad$ (The orbits are circles.),
- We can fix the phase by setting, for example, $p_{2}=0$.
- (Then $u_{2}(0)=0$.)
- This leaves a one-parameter family of periodic solutions.
(See Figures 33 and 34.).
- For nonlinear problems the family is generally not "vertical".

EXERCISE. (Demo lhb .)
Use AUTO to compute the zero stationary solution family, a Hopf bifurcation, and the emanating family of periodic solutions, of the "linear" Hopf bifurcation problem, i.e., of

$$
\left\{\begin{array}{l}
u_{1}^{\prime}=\lambda u_{1}-u_{2}  \tag{7}\\
u_{2}^{\prime}=u_{1}
\end{array}\right.
$$

NOTE:

- The family of periodic solutions is "vertical", i.e., $\lambda=0$ along it.
- This is not "typical" (not "generic") for Hopf bifurcation.
- For the numerically computed family, $\lambda=0$ up to numerical accuracy.
- The period is constant, namely, $2 \pi$, along the family.
- This is also not generic for periodic solutions from a Hopf bifurcation.
- The numerical computation of periodic solutions will be considered later.


Figure 33: Bifurcation diagram of the "linear" Hopf bifurcation problem.


Figure 34: A phase plot of some periodic solutions.

## The Hopf Bifurcation Theorem

THEOREM. Suppose that along a stationary solution family $(\mathbf{u}(\lambda), \lambda)$, of

$$
\mathbf{u}^{\prime}=\mathbf{f}(\mathbf{u}, \lambda)
$$

a complex conjugate pair of eigenvalues

$$
\alpha(\lambda) \pm i \beta(\lambda)
$$

of $f_{\mathbf{u}}(\mathbf{u}(\lambda), \lambda)$ crosses the imaginary axis transversally, i.e., for some $\lambda_{0}$,

$$
\alpha\left(\lambda_{0}\right)=0, \quad \beta\left(\lambda_{0}\right) \neq 0, \quad \text { and } \quad \dot{\alpha}\left(\lambda_{0}\right) \neq 0
$$

Also assume that there are no other eigenvalues on the imaginary axis.
Then there is a Hopf bifurcation, i.e., a family of periodic solutions bifurcates from the stationary solution at $\left(\mathbf{u}_{0}, \lambda_{0}\right)$.

NOTE: The assumptions also imply that $\mathbf{f}_{\mathbf{u}}^{0}$ is nonsingular, so that the stationary solution family can indeed be parametrized locally using $\lambda$.

EXERCISE. (AUTO Demo vhb.)
Use AUTO to compute the zero stationary solution family, a Hopf bifurcation, and the emanating family of periodic solutions for the equation

$$
\left\{\begin{array}{l}
u_{1}^{\prime}=\lambda u_{1}-u_{2},  \tag{8}\\
u_{2}^{\prime}=u_{1}\left(1-u_{1}\right) .
\end{array}\right.
$$

NOTE:

- $\mathbf{u}(t) \equiv \mathbf{0}$ is a stationary solution for all $\lambda$.
- $\mathbf{u}(t) \equiv\binom{1}{\lambda}$ is another stationary solution.

NOTE:
The Jacobian along the solution family $\mathbf{u}(t) \equiv \mathbf{0}$ is

$$
\left(\begin{array}{cc}
\lambda & -1 \\
1 & 0
\end{array}\right),
$$

with eigenvalues

$$
\frac{\lambda \pm \sqrt{\lambda^{2}-4}}{2}
$$

- The eigenvalues are complex for $\lambda \in(-2,2)$.
- The eigenvalues cross the imaginary axis when $\lambda$ passes through zero.
- Thus there is a Hopf bifurcation along $\mathbf{u} \equiv \mathbf{0}$ at $\lambda=0$.
- A family of periodic solutions bifurcates from $\mathbf{u}=\mathbf{0}$ at $\lambda=0$.


Figure 35: Bifurcation diagram for Equation (8).


Figure 36: A phase plot of some periodic solutions to Equation (8).


Figure 37: $u_{1}$ as a function of the scaled time variable $t$ for Equation (8).


Figure 38: $u_{2}$ as a function of the scaled time variable $t$ for Equation (8).

NOTE:

- The family of periodic solutions is also "vertical" (non-generic).
- The period changes along this family; in fact, the period tends to infinity.
- The terminating infinite period orbit is an example of a homoclinic orbit.
- This homoclinic orbit contains the stationary point $\left(u_{1}, u_{2}\right)=(1,0)$.
- In the solution diagrams, showing $u_{1}$ and $u_{2}$ versus time $t$, note how the "peak" in the solution remains in the same location.
- This is a result of the numerical "phase-condition", to be discussed later.

EXERCISE. (Demo het .)
Use AUTO to compute the zero stationary solution family, a Hopf bifurcation, and the emanating family of periodic solutions, of the equation

$$
\left\{\begin{array}{l}
u_{1}^{\prime}=\lambda u_{1}-u_{2}  \tag{9}\\
u_{2}^{\prime}=u_{1}\left(1-u_{1}^{2}\right)
\end{array}\right.
$$

NOTE:

- $\mathbf{u}(t) \equiv \mathbf{0}$ is a stationary solution for all $\lambda$.
- There is a Hopf bifurcation along $\mathbf{u} \equiv \mathbf{0}$ at $\lambda=0$.
- $\mathbf{u}(t) \equiv\binom{1}{\lambda}, \quad\binom{-1}{\lambda}$ are two more stationary solutions .


Figure 39: Bifurcation diagram for Equation (9).


Figure 40: A phase plot of some periodic solutions to Equation (9).


Figure 41: $u_{1}$ as a function of the scaled time variable $t$ for Equation (9).


Figure 42: $\quad u_{2}$ as a function of the scaled time variable $t$ for Equation (9).

- This family of periodic solutions is also "vertical" (non-generic).
- The period along this family also tends to infinity.
- The terminating infinite period orbit is an example of a heteroclinic cycle.
- This heteroclinic cycle is made up of two heteroclinic orbits.
- The heteroclinic orbits contains the stationary points

$$
\left(u_{1}, u_{2}\right)=(1,0) \quad \text { and } \quad\left(u_{1}, u_{2}\right)=(-1,0)
$$

EXERCISE. (AUTO demo pp3 .)
Compute the families of periodic solutions that bifurcate from the four Hopf bifurcation points in the following system; taking $p_{4}=4$.

$$
\begin{aligned}
u_{1}^{\prime}(t) & =u_{1}\left(1-u_{1}\right)-p_{4} u_{1} u_{2} \\
u_{2}^{\prime}(t) & =-\frac{1}{4} u_{2}+p_{4} u_{1} u_{2}-3 u_{2} u_{3}-p_{1}\left(1-e^{-5 u_{2}}\right) \\
u_{3}^{\prime}(t) & =-\frac{1}{2} u_{3}+3 u_{2} u_{3}
\end{aligned}
$$

This is a simple predator-prey model where, say,

$$
u_{1}=\text { plankton }, \quad u_{2}=\text { fish }, \quad u_{3}=\text { sharks }, \quad \lambda \equiv p_{1}=\text { fishing quota } .
$$

The factor

$$
\left(1-e^{-5 u_{2}}\right),
$$

models that the quota cannot be met if the fish population is small.


Figure 43: A bifurcation diagram for the 3 -species model; with $p_{4}=4$.

NOTE:

- These periodic solution families are not "vertical". (The generic case.)
- One family connects the two Hopf points along the stationary family along which $u_{1}$ is constant.
- The second family connects the two Hopf points along the stationary family along which $u_{1}$ is not constant.
- These two families of periodic solutions "intersect" at a branch point of periodic solutions, at $\lambda \approx 0.3012$.
- At this point there is an "interchange of stability" between the families.
- Stable periodic orbits are denoted by solid circles in the diagram.
- Unstable periodic orbits are denoted by open circles in the diagram.


Figure 44: Part of the planar orbit family for the 3 -species model; $p_{4}=4$.


Figure 45: The remainder of the planar orbit family; $p_{4}=4$.


Figure 46: Part of the 3D orbit family for the 3 -species model; $p_{4}=4$.


Figure 47: The remainder of the 3D orbit family; $p_{4}=4$.

## Computing Periodic Solutions

Periodic solutions can be computed very effectively using a boundary value approach. This method also determines the period very accurately. Moreover, the technique allows asymptotically unstable periodic orbits to be computed as easily as asymptotically stable ones.

## The BVP Approach.

Consider

$$
\mathbf{u}^{\prime}(t)=\mathbf{f}(\mathbf{u}(t), \lambda), \quad \mathbf{u}(\cdot), \mathbf{f}(\cdot) \in \mathrm{R}^{n}, \quad \lambda \in \mathrm{R}
$$

Fix the interval of periodicity by the transformation

$$
t \rightarrow \frac{t}{T}
$$

Then the equation becomes

$$
\mathbf{u}^{\prime}(t)=T \mathbf{f}(\mathbf{u}(t), \lambda), \quad \mathbf{u}(\cdot), \mathbf{f}(\cdot) \in \mathrm{R}^{n}, \quad T, \lambda \in \mathrm{R}
$$

and we seek solutions of period 1 , i.e.,

$$
\mathbf{u}(0)=\mathbf{u}(1)
$$

Note that the period $T$ is one of the unknowns.

The above equations do not uniquely specify $\mathbf{u}$ and $T$ :
Assume that we have computed

$$
\left(\mathbf{u}_{k-1}(\cdot), T_{k-1}, \lambda_{k-1}\right),
$$

and we want to compute the next solution

$$
\left(\mathbf{u}_{k}(\cdot), T_{k}, \lambda_{k}\right)
$$

Specifically, $\mathbf{u}_{k}(t)$ can be translated freely in time:
If $\mathbf{u}_{k}(t)$ is a periodic solution, then so is

$$
\mathbf{u}_{k}(t+\sigma),
$$

for any $\sigma$.
Thus, a "phase condition" is needed.

An example is the Poincaré orthogonality condition

$$
\left(\mathbf{u}_{k}(0)-\mathbf{u}_{k-1}(0)\right)^{*} \mathbf{u}_{k-1}^{\prime}(0)=0 .
$$

(Below we derive a numerically more suitable phase condition.)


Figure 48: Graphical interpretation of the Poincaré phase condition.

## Integral Phase Condition

If $\tilde{\mathbf{u}}_{k}(t)$ is a solution then so is

$$
\tilde{\mathbf{u}}_{k}(t+\sigma)
$$

for any $\sigma$.

We want the solution that minimizes

$$
D(\sigma) \equiv \int_{0}^{1}\left\|\tilde{\mathbf{u}}_{k}(t+\sigma)-\mathbf{u}_{k-1}(t)\right\|_{2}^{2} d t
$$

The optimal solution

$$
\tilde{\mathbf{u}}_{k}(t+\hat{\sigma}),
$$

must satisfy the necessary condition

$$
D^{\prime}(\hat{\sigma})=0 .
$$

Differentiation gives the necessary condition

$$
\int_{0}^{1}\left(\tilde{\mathbf{u}}_{k}(t+\hat{\sigma})-\mathbf{u}_{k-1}(t)\right)^{*} \tilde{\mathbf{u}}_{k}^{\prime}(t+\hat{\sigma}) d t=0 .
$$

Writing

$$
\mathbf{u}_{k}(t) \equiv \tilde{\mathbf{u}}_{k}(t+\hat{\sigma})
$$

gives

$$
\int_{0}^{1}\left(\mathbf{u}_{k}(t)-\mathbf{u}_{k-1}(t)\right)^{*} \mathbf{u}_{k}^{\prime}(t) d t=0 .
$$

Integration by parts, using periodicity, gives

$$
\int_{0}^{1} \mathbf{u}_{k}(t)^{*} \mathbf{u}_{k-1}^{\prime}(t) d t=0
$$

This is the integral phase condition.

## Pseudo-Arclength Continuation

We use pseudo-arclength continuation to follow a family of periodic solutions.

This allows calculation past folds along a family of periodic solutions.

It also allows calculation of a "vertical family" of periodic solutions.

For periodic solutions the pseudo-arclength equation is
$\int_{0}^{1}\left(\mathbf{u}_{k}(t)-\mathbf{u}_{k-1}(t)\right)^{*} \dot{\mathbf{u}}_{k-1}(t) d t+\left(T_{k}-T_{k-1}\right) \dot{T}_{k-1}+\left(\lambda_{k}-\lambda_{k-1}\right) \dot{\lambda}_{k-1}=\Delta s$.

In summary, we have the following equations for continuing periodic solutions:

$$
\begin{gathered}
\mathbf{u}_{k}^{\prime}(t)=T \mathbf{f}\left(\mathbf{u}_{k}(t), \lambda_{k}\right) \\
\mathbf{u}_{k}(0)=\mathbf{u}_{k}(1) \\
\int_{0}^{1} \mathbf{u}_{k}(t)^{*} \mathbf{u}_{k-1}^{\prime}(t) d t=0
\end{gathered}
$$

with pseudo-arclength continuation equation

$$
\int_{0}^{1}\left(\mathbf{u}_{k}(t)-\mathbf{u}_{k-1}(t)\right)^{*} \dot{\mathbf{u}}_{k-1}(t) d t+\left(T_{k}-T_{k-1}\right) \dot{T}_{k-1}+\left(\lambda_{k}-\lambda_{k-1}\right) \dot{\lambda}_{k-1}=\Delta s .
$$

Here

$$
\mathbf{u}(\cdot), \mathbf{f}(\cdot) \in \mathrm{R}^{n}, \quad \lambda, T \in \mathrm{R} .
$$

## Starting at a Hopf Bifurcation

Let

$$
\left(\mathbf{u}_{0}, \lambda_{0}\right)
$$

be a Hopf bifurcation point, i.e.,

$$
\mathbf{f}_{\mathbf{u}}\left(\mathbf{u}_{0}, \lambda_{0}\right)
$$

has a simple conjugate pair of purely imaginary eigenvalues

$$
\pm i \omega_{0}, \quad \omega_{0} \neq 0
$$

and no other eigenvalues on the imaginary axis.

Also, the pair crosses the imaginary axis transversally with respect to $\lambda$.

By the Hopf Bifurcation Theorem, a family of periodic solutions bifurcates.

Asymptotic estimates for periodic solutions near the Hopf bifurcation :

$$
\begin{aligned}
\mathbf{u}(t ; \epsilon) & =\mathbf{u}_{0}+\epsilon \boldsymbol{\phi}(t)+\mathcal{O}\left(\epsilon^{2}\right), \\
T(\epsilon) & =T_{0}+\mathcal{O}\left(\epsilon^{2}\right), \\
\lambda(\epsilon) & =\lambda_{0}+\mathcal{O}\left(\epsilon^{2}\right) .
\end{aligned}
$$

Here $\epsilon$ locally parametrizes the family of periodic solutions.
$T(\epsilon)$ denotes the period, and

$$
T_{0}=\frac{2 \pi}{\omega_{0}} .
$$

The function $\boldsymbol{\phi}(t)$ is the normalized nonzero periodic solution of the linearized, constant coefficient problem

$$
\boldsymbol{\phi}^{\prime}(t)=\mathrm{f}_{\mathbf{u}}\left(\mathbf{u}_{0}, \lambda_{0}\right) \boldsymbol{\phi}(t) .
$$

To compute a first periodic solution

$$
\left(\mathbf{u}_{1}(\cdot), T_{1}, \lambda_{1}\right),
$$

near a Hopf bifurcation $\left(\mathbf{u}_{0}, \lambda_{0}\right)$, we still have

$$
\begin{equation*}
\mathbf{u}_{1}^{\prime}(t)=T \mathbf{f}\left(\mathbf{u}_{1}(t), \lambda_{1}\right), \tag{10}
\end{equation*}
$$

$$
\begin{equation*}
\mathbf{u}_{1}(0)=\mathbf{u}_{1}(1) \text {. } \tag{11}
\end{equation*}
$$

Initial estimates for Newton's method are

$$
\mathbf{u}_{1}^{(0)}(t)=\mathbf{u}_{0}+\Delta s \boldsymbol{\phi}(t), \quad T_{1}^{(0)}=T_{0}, \quad \lambda_{1}^{(0)}=\lambda_{0} .
$$

Above, $\boldsymbol{\phi}(t)$ is a nonzero solution of the time-scaled, linearized equations

$$
\phi^{\prime}(t)=T_{0} \mathbf{f}_{\mathbf{u}}\left(\mathbf{u}_{0}, \lambda_{0}\right) \boldsymbol{\phi}(t), \quad \phi(0)=\boldsymbol{\phi}(1),
$$

namely,

$$
\boldsymbol{\phi}(t)=\sin (2 \pi t) \mathbf{w}_{s}+\cos (2 \pi t) \mathbf{w}_{c},
$$

where

$$
\left(\mathbf{w}_{s}, \mathbf{w}_{c}\right),
$$

is a null vector in

$$
\left(\begin{array}{cc}
-\omega_{0} I & \mathbf{f}_{\mathbf{u}}\left(\mathbf{u}_{0}, \lambda_{0}\right) \\
\mathbf{f}_{\mathbf{u}}\left(\mathbf{u}_{0}, \lambda_{0}\right) & \omega_{0} I
\end{array}\right)\binom{\mathbf{w}_{s}}{\mathbf{w}_{c}}=\binom{0}{\mathbf{0}}, \quad \omega_{0}=\frac{2 \pi}{T_{0}} .
$$

The nullspace is generically two-dimensional since

$$
\binom{-\mathbf{w}_{c}}{\mathbf{w}_{s}},
$$

is also a null vector.

For the phase equation we "align" $\mathbf{u}_{1}$ with $\boldsymbol{\phi}(t)$, i.e.,

$$
\int_{0}^{1} \mathbf{u}_{1}(t)^{*} \boldsymbol{\phi}^{\prime}(t) d t=0 \text {. }
$$

Since

$$
\dot{\lambda}_{0}=\dot{T}_{0}=0,
$$

the pseudo-arclength equation for the first step reduces to

$$
\int_{0}^{1}\left(\mathbf{u}_{1}(t)-\mathbf{u}_{0}(t)\right)^{*} \boldsymbol{\phi}(t) d t=\Delta s \text {. }
$$

## Accuracy Test

EXERCISE.

A simple accuracy test is to treat the linear equation

$$
\begin{aligned}
& u_{1}^{\prime}(t)=\lambda u_{1}-u_{2} \\
& u_{2}^{\prime}(t)=u_{1}
\end{aligned}
$$

as a bifurcation problem.

- It has a Hopf bifurcation point at $\lambda=0$ from the zero solution family.
- The bifurcating family of periodic solutions is vertical.
- Along the family, the period remains constant, namely, $T=2 \pi$.

For the above problem:

- Compute the family of periodic solutions, for different choices of the number of mesh points and the number of collocation points.
- Determine the error in the period for the computed solutions.

Typical results are shown in Table 1.

| ntst | ncol $=2$ |  | ncol $=3$ |  | ncol $=4$ |  |
| :---: | :---: | :---: | :--- | :--- | :--- | :--- |
| 4 | $0.47 \mathrm{e}-1$ | $(5.5)$ | $0.85 \mathrm{e}-3$ | $(7.9)$ | $0.85 \mathrm{e}-5$ | $(8.9)$ |
| 8 | $0.32 \mathrm{e}-2$ | $(4.0)$ | $0.14 \mathrm{e}-4$ | $(6.0)$ | $0.35 \mathrm{e}-7$ | $(8.0)$ |
| 16 | $0.20 \mathrm{e}-3$ | $(4.0)$ | $0.22 \mathrm{e}-6$ | $(6.0)$ | $0.14 \mathrm{e}-9$ | $(8.0)$ |
| 32 | $0.13 \mathrm{e}-4$ |  | $0.35 \mathrm{e}-8$ |  | $0.54 \mathrm{e}-11$ |  |

Table 1: Accuracy of T for the linear problem

EXAMPLE: (AUTO demo fhn.)

The FitzHugh-Nagumo model

$$
\begin{aligned}
& u_{1}^{\prime}=u_{1}-\frac{u_{1}^{3}}{3}-u_{2}+I \\
& u_{2}^{\prime}=a\left(u_{1}+b-c u_{2}\right)
\end{aligned}
$$

is a model of spike generation in squid giant axons, where
$u_{1}$ is the membrane potential,
$u_{2}$ is a recovery variable,
$I$ is the stimulus current.
Take $I$ as bifurcation parameter, and

$$
a=0.08 \quad, \quad b=0.7 \quad, \quad c=0.8
$$

If $I=0$ then $\left(u_{1}, u_{2}\right)=(-1.19941,-0.62426)$ is a stationary solution.


Figure 49: Bifurcation diagram of the Fitzhugh-Nagumo equations.


Figure 50: A stable periodic solutions .

## Periodically Forced Systems

Here we illustrate computing periodic solutions to a periodically forced system.
In AUTO this can be done by adding a nonlinear oscillator with the desired periodic forcing as one of its solution components.

An example of such an oscillator is

$$
\begin{aligned}
& x^{\prime}=x+\beta y-x\left(x^{2}+y^{2}\right) \\
& y^{\prime}=-\beta x+y-y\left(x^{2}+y^{2}\right)
\end{aligned}
$$

which has the asymptotically stable solution

$$
x=\sin (\beta t), \quad y=\cos (\beta t)
$$

EXAMPLE. (AUTO demo ffn.)
Couple the oscillator

$$
\begin{aligned}
x^{\prime} & =x+\beta y-x\left(x^{2}+y^{2}\right), \\
y^{\prime} & =-\beta x+y-y\left(x^{2}+y^{2}\right),
\end{aligned}
$$

to the Fitzhugh-Nagumo equations :

$$
\begin{aligned}
v^{\prime} & =c\left(v-\frac{v^{3}}{3}+w-r y\right) \\
w^{\prime} & =-(v-a+b * w) / c
\end{aligned}
$$

where

$$
b=0.8, \quad c=3, \quad \text { and } \quad \beta=10
$$

Note that if

$$
a=0 \quad \text { and } \quad r=0
$$

then a solution is

$$
x=\sin (\beta t), \quad y=\cos (\beta t), \quad v(t) \equiv 0, \quad w(t) \equiv 0
$$

Continue this solution from $r=0$ to, say, $r=10$.


Figure 51: Continuation from $r=0$ to $r=10$. Solution 1 is unstable. Solution 2 corresponds to a torus bifurcation.


Figure 52: Some solutions along the path from $r=0$ to $r=10$.

NOTE:

- The starting solution at $r=0$, with $v=w=0$, is unstable.
- The oscillation becomes stable when $r$ passes the value $r_{T} \approx 4.52$.
- At $r=r_{T}$ there is a torus bifurcation.


## General Non-Autonomous Systems

If the forcing is not periodic, or difficult to model by an autonomous oscillator, then the equations can be rewritten in autonomous form as follows:

$$
\begin{array}{ll}
\mathbf{u}^{\prime}(t)=\mathbf{f}(t, \mathbf{u}(t)), & \mathbf{u}(\cdot), \mathbf{f}(\cdot) \in \mathrm{R}^{n}, \quad t \in[0,1], \\
\mathbf{b}(\mathbf{u}(0), \mathbf{u}(1))=\mathbf{0}, & \mathbf{b}(\cdot) \in \mathrm{R}^{n},
\end{array}
$$

can be transformed into

$$
\begin{aligned}
& \mathbf{u}^{\prime}(t)=\mathbf{f}(\mathbf{v}(t), \mathbf{u}(t)), \\
& v^{\prime}(t)=1, \\
& \mathbf{b}(\mathbf{u}(0), \mathbf{u}(1))=\mathbf{0}, \\
& v(0)=0
\end{aligned}
$$

which is autonomous, with $n+1$ ODEs and $n+1$ boundary conditions.

## Periodic Solutions of Conservative Systems

EXAMPLE:

$$
\begin{aligned}
& u_{1}^{\prime}=-u_{2} \\
& u_{2}^{\prime}=u_{1}\left(1-u_{1}\right) .
\end{aligned}
$$

## PROBLEM:

- This equation has a family of periodic solutions, but no parameter !
- This system has a constant of motion, namely the Hamiltonian

$$
H\left(u_{1}, u_{2}\right)=-\frac{1}{2} u_{1}^{2}-\frac{1}{2} u_{2}^{2}+\frac{1}{3} u_{1}^{3} .
$$

## REMEDY:

Introduce an "unfolding term" with "unfolding parameter" $\lambda$ :

$$
\begin{aligned}
& u_{1}^{\prime}=\lambda u_{1}-u_{2}, \\
& u_{2}^{\prime}=u_{1}\left(1-u_{1}\right) .
\end{aligned}
$$

Then there is a "vertical" Hopf bifurcation from the trivial solution at $\lambda=0$.
(This is AUTO demo lhb; see Figures 33 and 34.)


Figure 53: Bifurcation diagram of the "linear" Hopf bifurcation problem.

NOTE:

- The family of periodic solutions is "vertical".
- The parameter $\lambda$ is solved for in each continuation step.
- Upon solving, $\lambda$ is found to be zero, up to numerical precision.
- One can use "standard" BVP continuation and bifurcation software.

EXAMPLE : The Circular Restricted 3-Body Problem (CR3BP).

$$
\begin{aligned}
x^{\prime \prime} & =2 y^{\prime}+x-\frac{(1-\mu)(x+\mu)}{r_{1}^{3}}-\frac{\mu(x-1+\mu)}{r_{2}^{3}} \\
y^{\prime \prime} & =-2 x^{\prime}+y-\frac{(1-\mu) y}{r_{1}^{3}}-\frac{\mu y}{r_{2}^{3}} \\
z^{\prime \prime} & =-\frac{(1-\mu) z}{r_{1}^{3}}-\frac{\mu z}{r_{2}^{3}}
\end{aligned}
$$

where

$$
r_{1}=\sqrt{(x+\mu)^{2}+y^{2}+z^{2}}, \quad r_{2}=\sqrt{(x-1+\mu)^{2}+y^{2}+z^{2}}
$$

and

$$
(x, y, z)
$$

denotes the position of the zero-mass body.
For the Earth-Moon system $\mu \approx 0.01215$.

The CR3BP has one integral of motion, namely, the "Jacobi-constant":

$$
J=\frac{x^{\prime 2}+y^{\prime 2}+z^{\prime 2}}{2}-U(x, y, z)-\mu \frac{1-\mu}{2},
$$

where

$$
U=\frac{1}{2}\left(x^{2}+y^{2}\right)+\frac{1-\mu}{r_{1}}+\frac{\mu}{r_{2}},
$$

where

$$
r_{1}=\sqrt{(x+\mu)^{2}+y^{2}+z^{2}}, \quad r_{2}=\sqrt{(x-1+\mu)^{2}+y^{2}+z^{2}} .
$$

## BOUNDARY VALUE FORMULATION:

$$
\begin{aligned}
x^{\prime} & =T v_{x} \\
y^{\prime} & =T v_{y}, \\
z^{\prime} & =T v_{z}, \\
v_{x}^{\prime} & =T\left[2 v_{y}+x-(1-\mu)(x+\mu) r_{1}^{-3}-\mu(x-1+\mu) r_{2}^{-3}+\lambda v_{x}\right], \\
v_{y}^{\prime} & =T\left[-2 v_{x}+y-(1-\mu) y r_{1}^{-3}-\mu y r_{2}^{-3}+\lambda v_{y}\right], \\
v_{z}^{\prime} & =T\left[-(1-\mu) z r_{1}^{-3}-\mu z r_{2}^{-3}+\lambda v_{z}\right],
\end{aligned}
$$

with periodicity boundary conditions

$$
\begin{gathered}
x(1)=x(0) \quad, \quad y(1)=y(0) \quad, \quad z(1)=z(0) \\
v_{x}(1)=v_{x}(0), \quad v_{y}(1)=v_{y}(0) \quad, \quad v_{z}(1)=v_{z}(0)
\end{gathered}
$$

+ phase constraint + pseudo-arclength equation.

NOTE:

- One can use standard BVP continuation and bifurcation software.
- The "unfolding term" $\lambda \nabla v$ regularizes the continuation.
- $\lambda$ will be "zero", once solved for.
- Other unfolding terms are possible.


Families of Periodic Solutions of the Earth-Moon system.


The planar Lyapunov family L1.


The Halo family H1.


The Halo family H1.


The Vertical family V1.


The Axial family A1.

## Following Periodic Orbit Folds

Fold-following algorithms also apply to folds along solution families of boundary value problems and, in particular, folds along families of periodic solutions.

EXAMPLE: The $A \rightarrow B \rightarrow C$ reaction. (AUTO demo abc.)

The equations are

$$
\begin{aligned}
u_{1}^{\prime} & =-u_{1}+D\left(1-u_{1}\right) e^{u_{3}} \\
u_{2}^{\prime} & =-u_{2}+D\left(1-u_{1}\right) e^{u_{3}}-D \sigma u_{2} e^{u_{3}} \\
u_{3}^{\prime} & =-u_{3}-\beta u_{3}+D B\left(1-u_{1}\right) e^{u_{3}}+D B \alpha \sigma u_{2} e^{u_{3}}
\end{aligned}
$$

with

$$
\alpha=1 \quad, \quad \sigma=0.04 \quad, \quad B=8
$$

We will compute solutions for varying $D$ and $\beta$.


Figure 54: Stationary and periodic solutions of demo abc; $\beta=1.55$.

Recall that periodic orbits families can be computed using the equations

$$
\begin{gathered}
\mathbf{u}^{\prime}(t)-T \mathbf{f}(\mathbf{u}(t), \lambda)=\mathbf{0}, \\
\mathbf{u}(0)-\mathbf{u}(1)=\mathbf{0}, \\
\int_{0}^{1} \mathbf{u}(t)^{*} \mathbf{u}_{0}^{\prime}(t) d t=0
\end{gathered}
$$

where $\mathbf{u}_{0}$ is a reference orbit, typically the latest computed orbit.

The above boundary value problem is of the form

$$
\mathbf{F}(\mathbf{X}, \lambda)=\mathbf{0}
$$

where

$$
\mathbf{X}=(\mathbf{u}, T)
$$

At a fold with respect to $\lambda$ we have

$$
\begin{aligned}
\mathbf{F}_{\mathbf{X}}(\mathbf{X}, \lambda) \boldsymbol{\Phi} & =\mathbf{0}, \\
\langle\boldsymbol{\Phi}, \boldsymbol{\Phi}\rangle & =1
\end{aligned}
$$

where

$$
\mathbf{X}=(\mathbf{u}, T) \quad, \quad \mathbf{\Phi}=(\mathbf{v}, S)
$$

or, written in detail,

$$
\begin{gathered}
\mathbf{v}^{\prime}(t)-T \mathbf{f}_{\mathbf{u}}(\mathbf{u}(t), \lambda) \mathbf{v}-S \mathbf{f}(\mathbf{u}(t), \lambda)=\mathbf{0}, \\
\mathbf{v}(0)-\mathbf{v}(1)=\mathbf{0}, \\
\int_{0}^{1} \mathbf{v}(t)^{*} \mathbf{u}_{0}^{\prime}(t) d t=0, \\
\int_{0}^{1} \mathbf{v}(t)^{*} \mathbf{v}(t) d t+S^{2}=1
\end{gathered}
$$

The complete extended system to follow a fold is

$$
\begin{gathered}
\mathbf{F}(\mathbf{X}, \lambda, \mu)=\mathbf{0} \\
\mathbf{F}_{\mathbf{X}}(\mathbf{X}, \lambda, \mu) \boldsymbol{\Phi}=\mathbf{0} \\
<\mathbf{\Phi}, \boldsymbol{\Phi}>-1=0
\end{gathered}
$$

with two free problem parameters $\lambda$ and $\mu$.

To the above we add the pseudo-arclength equation
$<\mathbf{X}-\mathbf{X}_{0}, \dot{\mathbf{X}}_{0}>+<\boldsymbol{\Phi}-\boldsymbol{\Phi}_{0}, \dot{\boldsymbol{\Phi}}_{0}>+\left(\lambda-\lambda_{0}\right) \dot{\lambda}_{0}+\left(\mu-\mu_{0}\right) \dot{\mu}_{0}-\Delta s=0$.

In detail:

$$
\begin{gathered}
\mathbf{u}^{\prime}(t)-T \mathbf{f}(\mathbf{u}(t), \lambda, \mu)=\mathbf{0} \\
\mathbf{u}(0)-\mathbf{u}(1)=\mathbf{0} \\
\int_{0}^{1} \mathbf{u}(t)^{*} \mathbf{u}_{0}^{\prime}(t) d t=0 \\
\mathbf{v}^{\prime}(t)-T \mathbf{f}_{\mathbf{u}}(\mathbf{u}(t), \lambda, \mu) \mathbf{v}-S \mathbf{f}(\mathbf{u}(t), \lambda, \mu)=\mathbf{0}, \\
\mathbf{v}(0)-\mathbf{v}(1)=\mathbf{0} \\
\int_{0}^{1} \mathbf{v}(t)^{*} \mathbf{u}_{0}^{\prime}(t) d t=0
\end{gathered}
$$

with normalization

$$
\int_{0}^{1} \mathbf{v}(t)^{*} \mathbf{v}(t) d t+S^{2}-1=0
$$

and pseudo-arclength equation

$$
\begin{aligned}
& \int_{0}^{1}\left(\mathbf{u}(t)-\mathbf{u}_{0}(t)\right)^{*} \dot{\mathbf{u}}_{0}(t) d t+\int_{0}^{1}\left(\mathbf{v}(t)-\mathbf{v}_{0}(t)\right)^{*} \dot{\mathbf{v}}_{0}(t) d t+ \\
& \quad+\left(T_{0}-T\right) \dot{T}_{0}+\left(S_{0}-S\right) \dot{S}_{0}+\left(\lambda-\lambda_{0}\right) \dot{\lambda}_{0}+\left(\mu-\mu_{0}\right) \dot{\mu}_{0}-\Delta s=0
\end{aligned}
$$



Figure 55: The locus of periodic solution folds of demo abc.


Figure 56: Stationary and periodic solutions of demo abc; $\beta=1.55$.


Figure 57: Stationary and periodic solutions of demo abc; $\beta=1.56$.


Figure 58: Stationary and periodic solutions of demo abc; $\beta=1.57$.


Figure 59: Stationary and periodic solutions of demo abc; $\beta=1.58$.


Figure 60: Stationary and periodic solutions of demo abc; $\beta=1.61$.


Figure 61: Stationary and periodic solutions of demo abc; $\beta=1.62$.

## Following Hopf Bifurcations

We consider the persistence of a Hopf bifurcation as a second parameter is varied, and we give an algorithm for computing a 2-parameter locus of Hopf bifurcation points.

A Hopf bifurcation along a stationary solution family $(\mathbf{u}(\lambda), \lambda)$, of

$$
\mathbf{u}^{\prime}=\mathbf{f}(\mathbf{u}, \lambda),
$$

occurs when a complex conjugate pair of eigenvalues

$$
\alpha(\lambda) \pm i \beta(\lambda),
$$

of $f_{\mathbf{u}}(\mathbf{u}(\lambda), \lambda)$ crosses the imaginary axis transversally, i.e., for some $\lambda_{0}$,

$$
\alpha\left(\lambda_{0}\right)=0, \quad \dot{\alpha}\left(\lambda_{0}\right) \neq 0, \quad \text { and } \quad \beta_{0}=\beta\left(\lambda_{0}\right) \neq 0,
$$

also assuming there are no other eigenvalues on the imaginary axis.
The assumptions imply that $\mathbf{f}_{u}^{0}$ is nonsingular, so that stationary solution family can indeed be parametrized locally using $\lambda$.

The right and left complex eigenvectors of $\mathbf{f}_{\mathbf{u}}^{0}=f_{\mathbf{u}}\left(\mathbf{u}\left(\lambda_{0}\right), \lambda_{0}\right)$ are defined by

$$
\mathbf{f}_{\mathbf{u}}^{0} \phi_{0}=i \beta_{0} \phi_{0} \quad, \quad\left(\mathbf{f}_{\mathbf{u}}^{0}\right)^{*} \boldsymbol{\psi}_{0}=-i \beta_{0} \boldsymbol{\psi}_{0}
$$

## Transversality and Persistence

THEOREM. The eigenvalue crossing in the Hopf Bifurcation Theorem is transversal if

$$
\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\mathbf{f}_{\mathbf{u u}}^{0}\left(\mathbf{f}_{\mathbf{u}}^{0}\right)^{-1} \mathbf{f}_{\lambda}^{0}-\mathbf{f}_{\mathbf{u} \lambda}^{0}\right] \boldsymbol{\phi}_{0}\right) \neq 0
$$

PROOF. Since the eigenvalue

$$
i \beta_{0}
$$

is algebraically simple, there is a smooth solution family (at least locally) to the parametrized right and left eigenvalue-eigenvector equations :
(a) $\quad \mathbf{f}_{\mathbf{u}}(\mathbf{u}(\lambda), \lambda) \boldsymbol{\phi}(\lambda)=\kappa(\lambda) \boldsymbol{\phi}(\lambda)$,
(b) $\quad \boldsymbol{\psi}(\lambda)^{*} \mathbf{f}_{\mathbf{u}}(\mathbf{u}(\lambda), \lambda)=\kappa(\lambda) \boldsymbol{\psi}^{*}(\lambda)$,
(c) $\quad \boldsymbol{\psi}(\lambda)^{*} \boldsymbol{\phi}(\lambda)=1, \quad$ (and also $\left.\boldsymbol{\phi}(\lambda)^{*} \boldsymbol{\phi}(\lambda)=1\right)$,
with

$$
\kappa\left(\lambda_{0}\right)=i \beta_{0} .
$$

(Above, * denotes conjugate transpose.)
$\mathbf{f}_{\mathbf{u}}(\mathbf{u}(\lambda), \lambda) \boldsymbol{\phi}(\lambda)=\kappa(\lambda) \boldsymbol{\phi}(\lambda) \quad, \quad \boldsymbol{\psi}(\lambda)^{*} \mathbf{f}_{\mathbf{u}}(\mathbf{u}(\lambda), \lambda)=\kappa(\lambda) \boldsymbol{\psi}^{*}(\lambda) \quad, \quad \boldsymbol{\psi}(\lambda)^{*} \boldsymbol{\phi}(\lambda)=1$

Differentiation with respect to $\lambda$ gives
(a) $\mathbf{f}_{\mathbf{u u}} \dot{\mathbf{u}} \boldsymbol{\phi}+\mathbf{f}_{\mathbf{u} \lambda} \boldsymbol{\phi}+\mathbf{f}_{\mathbf{u}} \dot{\boldsymbol{\phi}}=\dot{\kappa} \boldsymbol{\phi}+\kappa \dot{\boldsymbol{\phi}}$,
(b) $\quad \boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u u}} \dot{\mathbf{u}}+\boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u} \lambda}+\dot{\boldsymbol{\psi}}^{*} \mathbf{f}_{\mathbf{u}}=\dot{\kappa} \boldsymbol{\psi}^{*}+\kappa \dot{\boldsymbol{\psi}}^{*}$,
(c) $\dot{\boldsymbol{\psi}}^{*} \boldsymbol{\phi}+\boldsymbol{\psi}^{*} \dot{\boldsymbol{\phi}}=0$.

Multiply and
(a) on the left by $\boldsymbol{\psi}^{*}$,
(b) on the right by $\phi$,
to get
(a) $\quad \boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u u}} \dot{u} \boldsymbol{\phi}+\boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u} \lambda} \boldsymbol{\phi}+\boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u}} \dot{\boldsymbol{\phi}}=\dot{\kappa} \boldsymbol{\psi}^{*} \boldsymbol{\phi}+\kappa \boldsymbol{\psi}^{*} \dot{\boldsymbol{\phi}}$,
(b) $\quad \boldsymbol{\psi}^{*} \mathrm{f}_{\mathrm{uu}} \dot{u} \boldsymbol{\phi}+\boldsymbol{\psi}^{*} \mathrm{f}_{\mathrm{u} \lambda} \boldsymbol{\phi}+\dot{\boldsymbol{\psi}}^{*} \mathrm{f}_{\mathbf{u}} \boldsymbol{\phi}=\dot{\kappa} \boldsymbol{\psi}^{*} \boldsymbol{\phi}+\kappa \dot{\boldsymbol{\psi}}^{*} \boldsymbol{\phi}$.

$$
\begin{aligned}
& \psi^{*} \mathbf{f}_{\mathbf{u u}} \dot{u} \boldsymbol{\phi}+\boldsymbol{\psi}^{*} \mathrm{f}_{\mathbf{u} \lambda} \boldsymbol{\phi}+\boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u}} \dot{\boldsymbol{\phi}}=\dot{\kappa} \boldsymbol{\psi}^{*} \boldsymbol{\phi}+\kappa \boldsymbol{\psi}^{*} \dot{\phi} \\
& \boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u u}} \dot{u} \boldsymbol{\phi}+\boldsymbol{\psi}^{*} \mathrm{f}_{\mathbf{u} \lambda} \boldsymbol{\phi}+\dot{\boldsymbol{\psi}}^{*} \mathrm{f}_{\mathbf{u}} \boldsymbol{\phi}=\dot{\kappa} \boldsymbol{\psi}^{*} \boldsymbol{\phi}+\kappa \dot{\boldsymbol{\psi}}^{*} \boldsymbol{\phi}
\end{aligned}
$$

Adding the above, and using

$$
\underbrace{\boldsymbol{\psi}^{*} \mathbf{f}_{\mathbf{u}}}_{=\kappa \boldsymbol{\psi}^{*}} \dot{\boldsymbol{\phi}}+\dot{\boldsymbol{\psi}}^{*} \underbrace{\mathbf{f}_{u} \phi}_{=\kappa \boldsymbol{\kappa}}=\kappa\left(\boldsymbol{\psi}^{*} \dot{\boldsymbol{\phi}}+\dot{\boldsymbol{\psi}}^{*} \boldsymbol{\phi}\right)=\kappa \frac{d}{d \lambda}(\underbrace{\boldsymbol{\psi}^{*} \phi}_{=1})=0
$$

we find

$$
\dot{\kappa}=\boldsymbol{\psi}^{*}\left[\mathbf{f}_{\mathbf{u u}} \dot{\mathbf{u}}+\mathbf{f}_{\mathbf{u} \lambda}\right] \boldsymbol{\phi}
$$

$$
\dot{\kappa}=\psi^{*}\left[\mathbf{f}_{\mathbf{u u}} \dot{\mathbf{u}}+\mathbf{f}_{\mathbf{u} \lambda}\right] \phi
$$

From differentiating

$$
\mathbf{f}(\mathbf{u}(\lambda), \lambda)=\mathbf{0}
$$

with respect to $\lambda$, we have

$$
\dot{\mathbf{u}}=-\left(\mathbf{f}_{\mathbf{u}}\right)^{-1} \mathbf{f}_{\lambda},
$$

so that

$$
\dot{\kappa}=\boldsymbol{\psi}^{*}\left[-\mathbf{f}_{\mathbf{u u}}\left(\mathbf{f}_{\mathbf{u}}\right)^{-1} \mathbf{f}_{\lambda}+\mathbf{f}_{\mathbf{u} \lambda}\right] \phi
$$

Thus the eigenvalue crossing is transversal if

$$
\dot{\alpha}(0)=\operatorname{Re}\left(\dot{\kappa}_{0}\right)=\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\mathbf{f}_{\mathbf{u u}}^{0}\left(\mathbf{f}_{\mathbf{u}}^{0}\right)^{-1} \mathbf{f}_{\lambda}^{0}-\mathbf{f}_{\mathbf{u} \lambda}^{0}\right] \boldsymbol{\phi}_{0}\right) \neq 0
$$

NOTE:

The transversality condition of the Theorem, i.e.,

$$
\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\mathbf{f}_{\mathbf{u u}}^{0}\left(\mathbf{f}_{\mathbf{u}}^{0}\right)^{-1} \mathbf{f}_{\lambda}^{0}-\mathbf{f}_{\mathbf{u} \lambda}^{0}\right] \boldsymbol{\phi}_{0}\right) \neq 0,
$$

is also needed for persistence of the Hopf bifurcation, as discussed below.

The extended system for following Hopf bifurcations is

$$
\mathbf{F}(\mathbf{u}, \boldsymbol{\phi}, \beta, \lambda ; \mu) \equiv\left\{\begin{array}{l}
\mathbf{f}(\mathbf{u}, \lambda, \mu)=\mathbf{0} \\
\mathbf{f}_{\mathbf{u}}(\mathbf{u}, \lambda, \mu) \boldsymbol{\phi}-i \beta \boldsymbol{\phi}=\mathbf{0} \\
\boldsymbol{\phi}^{*} \boldsymbol{\phi}_{0}-1=0
\end{array}\right.
$$

where

$$
\mathbf{F}: \quad \mathrm{R}^{n} \times \mathrm{C}^{n} \times \mathrm{R}^{2} \times \mathrm{R} \quad \rightarrow \quad \mathrm{R}^{n} \times \mathrm{C}^{n} \times \mathrm{C},
$$

and to which we want to compute a solution family

$$
(\mathbf{u}, \boldsymbol{\phi}, \beta, \lambda, \mu)
$$

with

$$
\mathbf{u} \in \mathrm{R}^{n}, \quad \phi \in \mathrm{C}^{n}, \quad \beta, \lambda, \mu \in \mathrm{R} .
$$

Above $\phi_{0}$ belongs to a "reference solution"

$$
\left(\mathbf{u}_{0}, \phi_{0}, \beta_{0}, \lambda_{0}, \mu_{0}\right)
$$

which typically is the latest computed solution point of a family.

First consider parametrizing in the second parameter $\mu$, i.e., we seek a family

$$
(\mathbf{u}(\mu), \boldsymbol{\phi}(\mu), \beta(\mu), \lambda(\mu)) .
$$

(In practice, pseudo-arclength continuation is used.)
The derivative with respect to

$$
(\mathbf{u}, \phi, \beta, \lambda),
$$

at the solution point

$$
\left(\mathbf{u}_{0}, \phi_{0}, \beta_{0}, \lambda_{0}, \mu_{0}\right),
$$

is

$$
\left(\begin{array}{cccc}
\mathbf{f}_{\mathbf{u}}^{0} & O & \mathbf{0} & \mathbf{f}_{\lambda}^{0}  \tag{12}\\
\mathbf{f}_{\mathbf{u u}}^{0} \boldsymbol{\phi}_{0} & \mathbf{f}_{\mathbf{u}}^{0}-i \beta_{0} I & -i \boldsymbol{\phi}_{0} & \mathbf{f}_{\mathbf{u} \lambda}^{0} \boldsymbol{\phi}_{0} \\
\mathbf{0}^{*} & \boldsymbol{\phi}_{0}^{*} & 0 & 0
\end{array}\right) .
$$

The Jacobian is of the form

$$
\left(\begin{array}{cccc}
A & O & \mathbf{0} & c_{1} \\
C & D & -i \boldsymbol{\phi}_{0} & c_{2} \\
\mathbf{0}^{*} & \boldsymbol{\phi}_{0}^{*} & 0 & 0
\end{array}\right),
$$

where

$$
A=\mathbf{f}_{\mathbf{u}}^{0} \quad \text { (nonsingular) }, \quad C=\mathbf{f}_{\mathbf{u u}}^{0} \phi_{0}, \quad D=\mathbf{f}_{\mathbf{u}}^{0}-i \beta_{0} I,
$$

and

$$
c_{1}=\mathbf{f}_{\lambda}^{0}, \quad c_{2}=\mathbf{f}_{\mathbf{u} \lambda}^{0} \phi_{0},
$$

with

$$
\mathcal{N}(D)=\operatorname{Span}\left\{\boldsymbol{\phi}_{0}\right\}, \quad \mathcal{N}\left(D^{*}\right)=\operatorname{Span}\left\{\boldsymbol{\psi}_{0}\right\},
$$

where

$$
\boldsymbol{\psi}_{0}^{*} \phi_{0}=1, \quad \phi_{0}^{*} \phi_{0}=1
$$

## THEOREM.

If the eigenvalue crossing is transversal, i.e., if

$$
\operatorname{Re}\left(\dot{\kappa}_{0}\right) \neq 0
$$

then the Jacobian matrix (12) is nonsingular.

Hence there locally exists a solution family

$$
(\mathbf{u}(\mu), \boldsymbol{\phi}(\mu), \beta(\mu), \lambda(\mu))
$$

to the extended system

$$
\mathbf{F}(\mathbf{u}, \phi, \beta, \lambda ; \mu)=\mathbf{0},
$$

i.e., the Hopf bifurcation persists under small perturbations of $\mu$.

PROOF. We prove this by constructing a solution

$$
\mathbf{x} \in \mathrm{R}^{n}, \quad \mathbf{y} \in \mathrm{C}^{n}, \quad z_{1}, z_{2} \in \mathrm{R}
$$

to

$$
\left(\begin{array}{cccc}
A & O & \mathbf{0} & \mathbf{c}_{1} \\
C & D & -i \boldsymbol{\phi}_{0} & \mathbf{c}_{2} \\
\mathbf{0}^{*} & \boldsymbol{\phi}_{0}^{*} & 0 & 0
\end{array}\right)\left(\begin{array}{c}
\mathbf{x} \\
\mathbf{y} \\
z_{1} \\
z_{2}
\end{array}\right)=\left(\begin{array}{c}
\mathbf{f} \\
\mathbf{g} \\
h
\end{array}\right),
$$

where

$$
\mathbf{f} \in \mathrm{R}^{n}, \quad \mathbf{g} \in \mathrm{C}^{n}, \quad h \in \mathrm{C} .
$$

From the first equation

$$
A \mathbf{x}+z_{2} \mathbf{c}_{1}=\mathbf{f}
$$

we have

$$
\mathbf{x}=A^{-1} \mathbf{f}-z_{2} A^{-1} \mathbf{c}_{1} .
$$

The second equation can then be written

$$
C A^{-1} \mathbf{f}-z_{2} C A^{-1} \mathbf{c}_{1}+D \mathbf{y}-z_{1} i \phi_{0}+z_{2} \mathbf{c}_{2}=\mathbf{g} .
$$

$$
C A^{-1} \mathbf{f}-z_{2} C A^{-1} \mathbf{c}_{1}+D \mathbf{y}-z_{1} i \phi_{0}+z_{2} \mathbf{c}_{2}=\mathbf{g}
$$

Multiply on the left by $\boldsymbol{\psi}_{0}^{*}$ to get

$$
\boldsymbol{\psi}_{0}^{*} C A^{-1} \mathbf{f}-z_{2} \boldsymbol{\psi}_{0}^{*} C A^{-1} \mathbf{c}_{1}-z_{1} i \boldsymbol{\psi}_{0}^{*} \boldsymbol{\phi}_{0}+z_{2} \boldsymbol{\psi}_{0}^{*} \mathbf{c}_{2}=\boldsymbol{\psi}_{0}^{*} \mathbf{g}
$$

Recall that

$$
\boldsymbol{\psi}_{0}^{*} \phi_{0}=1
$$

Defining

$$
\tilde{\mathbf{f}} \equiv C A^{-1} \mathbf{f}, \quad \tilde{\mathbf{c}}_{1} \equiv C A^{-1} \mathbf{c}_{1}
$$

we have

$$
i z_{1}+\boldsymbol{\psi}_{0}^{*}\left(\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right) z_{2}=\boldsymbol{\psi}_{0}^{*}(\tilde{\mathbf{f}}-\mathbf{g}) .
$$

Computationally $\tilde{\mathbf{f}}$ and $\tilde{\mathbf{c}}_{1}$ are obtained from

$$
\begin{array}{ll}
A \hat{\mathbf{f}}=\mathbf{f} & , \begin{array}{l} 
\\
\hat{\mathbf{c}}_{1}=\mathbf{c}_{1} \\
\tilde{\mathbf{f}}=C \hat{\mathbf{f}}
\end{array} \\
& , \tilde{\mathbf{c}}_{1}=C \hat{\mathbf{c}}_{1}
\end{array}
$$

$$
i z_{1}+\boldsymbol{\psi}_{0}^{*}\left(\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right) z_{2}=\boldsymbol{\psi}_{0}^{*}(\tilde{\mathbf{f}}-\mathbf{g})
$$

Separate real and imaginary part of this equation, and use the fact that

$$
z_{1} \text { and } z_{2} \text { are real, }
$$

to get

$$
\begin{aligned}
\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right]\right) z_{2} & =\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}[\tilde{\mathbf{f}}-\mathbf{g}]\right), \\
z_{1}+\operatorname{Im}\left(\boldsymbol{\psi}_{0}^{*}\left[\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right]\right) z_{2} & =\operatorname{Im}\left(\boldsymbol{\psi}_{0}^{*}[\tilde{\mathbf{f}}-\mathbf{g}]\right),
\end{aligned}
$$

from which

$$
z_{2}=\frac{\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}[\tilde{\mathbf{f}}-\mathbf{g}]\right)}{\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right]\right)},
$$

$$
z_{1}=-z_{2} \operatorname{Im}\left(\boldsymbol{\psi}_{0}^{*}\left[\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right]\right)+\operatorname{Im}\left(\boldsymbol{\psi}_{0}^{*}[\tilde{\mathbf{f}}-\mathbf{g}]\right) .
$$

Now solve for $\mathbf{x}$ in

$$
A \mathbf{x}=\mathbf{f}-z_{2} \mathbf{c}_{1}
$$

and compute a particular solution $\mathbf{y}_{p}$ to

$$
D \mathbf{y}=\mathbf{g}-C \mathbf{x}+i z_{1} \phi_{0}-z_{2} \mathbf{c}_{2}
$$

Then

$$
\mathbf{y}=\mathbf{y}_{p}+\alpha \boldsymbol{\phi}_{0}, \quad \alpha \in \mathrm{C}
$$

The third equation is

$$
\phi_{0}^{*} \mathbf{y}=\boldsymbol{\phi}_{0}^{*} \mathbf{y}_{p}+\alpha \boldsymbol{\phi}_{0}^{*} \boldsymbol{\phi}_{0}=h
$$

from which, using $\phi_{0}^{*} \phi_{0}=1$,

$$
\alpha=h-\boldsymbol{\phi}_{0}^{*} \mathbf{y}_{p} .
$$

The above construction can be carried out if

$$
\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right]\right) \neq 0
$$

However, using the definition of $\tilde{\mathbf{c}}_{1}$ and $\mathbf{c}_{2}$, we have

$$
\begin{aligned}
\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\tilde{\mathbf{c}}_{1}-\mathbf{c}_{2}\right]\right) & =\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[C A^{-1} \mathbf{c}_{1}-\mathbf{c}_{2}\right]\right) \\
& =\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\mathbf{f}_{\mathbf{u u}}^{0} \boldsymbol{\phi}_{0}\left(\mathbf{f}_{\mathbf{u}}^{0}\right)^{-1} \mathbf{f}_{\lambda}^{0}-\mathbf{f}_{\mathbf{u} \lambda}^{0} \boldsymbol{\phi}_{0}\right]\right) \\
& =\operatorname{Re}\left(\boldsymbol{\psi}_{0}^{*}\left[\mathbf{f}_{\mathbf{u u}}^{0}\left(\mathbf{f}_{\mathbf{u}}^{0}\right)^{-1} \mathbf{f}_{\lambda}^{0}-\mathbf{f}_{\mathbf{u} \lambda}^{0}\right] \boldsymbol{\phi}_{0}\right) \\
& =\operatorname{Re}\left(\dot{\kappa}_{0}\right) \neq 0 . \bullet
\end{aligned}
$$

## Practical Continuation of Hopf Bifurcations

Recall that the extended system for following Hopf bifurcations is

$$
\mathbf{F}(\mathbf{u}, \boldsymbol{\phi}, \beta, \lambda ; \mu) \equiv\left\{\begin{array}{l}
\mathbf{f}(\mathbf{u}, \lambda, \mu)=\mathbf{0} \\
\mathbf{f}_{\mathbf{u}}(\mathbf{u}, \lambda, \mu) \boldsymbol{\phi}-i \beta \boldsymbol{\phi}=\mathbf{0} \\
\boldsymbol{\phi}^{*} \boldsymbol{\phi}_{0}-1=0
\end{array}\right.
$$

where

$$
\mathrm{F}: \quad \mathrm{R}^{n} \times \mathrm{C}^{n} \times \mathrm{R}^{2} \times \mathrm{R} \quad \rightarrow \quad \mathrm{R}^{n} \times \mathrm{C}^{n} \times \mathrm{C},
$$

and to which we want to compute a solution family

$$
(\mathbf{u}, \phi, \beta, \lambda, \mu), \quad \text { with } \mathbf{u} \in \mathrm{R}^{n}, \quad \phi \in \mathrm{C}^{n}, \quad \beta, \lambda, \mu \in \mathrm{R} .
$$

In practice, we treat $\mu$ as an unknown, and add the continuation equation $\left(\mathbf{u}-\mathbf{u}_{0}\right)^{*} \dot{\mathbf{u}}_{0}+\left(\boldsymbol{\phi}-\boldsymbol{\phi}_{0}\right)^{*} \dot{\phi}_{0}+\left(\beta-\beta_{0}\right) \dot{\beta}_{0}+\left(\lambda-\lambda_{0}\right) \dot{\lambda}_{0}+\left(\mu-\mu_{0}\right) \dot{\mu}_{0}-\Delta s=0$.

## EXERCISE.

Investigate the Hopf bifurcations in the system

$$
\begin{aligned}
& u_{1}^{\prime}(t)=u_{1}\left(1-u_{1}\right)-p_{4} u_{1} u_{2} \\
& u_{2}^{\prime}(t)=-\frac{1}{4} u_{2}+p_{4} u_{1} u_{2}-3 u_{2} u_{3}-p_{1}\left(1-e^{-5 u_{2}}\right), \\
& u_{3}^{\prime}(t)=-\frac{1}{2} u_{3}+3 u_{2} u_{3}
\end{aligned}
$$

This is the predator-prey model where,
$u_{1} \sim$ plankton, $\quad u_{2} \sim$ fish,$\quad u_{3} \sim$ sharks,$\quad \lambda \equiv p_{1} \sim$ fishing quota.
(See also Figure 43.)


Figure 62: A bifurcation diagram for the 3 -species model; with $p_{4}=3$.


Figure 63: Loci of Hopf bifurcations for the 3 -species model.


Figure 64: A bifurcation diagram for the 3 -species model; with $p_{4}=4$.

EXERCISE. (AUTO demo abc-hb .)

Compute loci of Hopf bifurcation points for the $A \rightarrow B \rightarrow C$ reaction

$$
\begin{aligned}
u_{1}^{\prime} & =-u_{1}+D\left(1-u_{1}\right) e^{u_{3}} \\
u_{2}^{\prime} & =-u_{2}+D\left(1-u_{1}\right) e^{u_{3}}-D \sigma u_{2} e^{u_{3}} \\
u_{3}^{\prime} & =-u_{3}-\beta u_{3}+D B\left(1-u_{1}\right) e^{u_{3}}+D B \alpha \sigma u_{2} e^{u_{3}}
\end{aligned}
$$

with

$$
\alpha=1 \quad, \quad \sigma=0.04 \quad, \quad B=8
$$

and varying $D$ and $\beta$.


Figure 65: A locus of Hopf bifurcations.


Figure 66: Diagrams for $\beta=1.2,1.3,1.4,1.5,1.6,1.7,1.8$.

## Stable and Unstable Manifolds

- One can also use continuation to compute solution families of IVP.
- In particular, one can compute stable and unstable manifolds .

EXAMPLE: The Lorenz Equations. (AUTO demos lor, lrz, man.)

$$
\begin{aligned}
x^{\prime} & =\sigma(y-x) \\
y^{\prime} & =\rho x-y-x z \\
z^{\prime} & =x y-\beta z,
\end{aligned}
$$

where

$$
\sigma=10 \quad \text { and } \quad \beta=8 / 3
$$

The Lorenz Equations


Figure 67: Bifurcation diagram of the Lorenz equations.

- The zero solution is unstable for $\rho>1$.
- Two nonzero stationary solutions bifurcate at $\rho=1$.
- The nonzero stationary solutions become unstable for $\rho>\rho_{H}$.
- At $\rho_{H}\left(\rho_{H} \approx 24.7\right)$ there are Hopf bifurcations.
- Unstable periodic solutions emanate from each Hopf bifurcation.
- These families end in homoclinic orbits (infinite period) at $\rho \approx 13.9$.
- For $\rho>\rho_{H}$ there is the famous Lorenz attractor.

The Lorenz Equations


Figure 68: Unstable periodic orbits of the Lorenz equations.

## The Lorenz Manifold

- For $\rho>1$ the origin is a saddle point.
- The Jacobian has two negative eigenvalues and one positive eigenvalue.
- The two negative eigenvalues give rise to a 2D stable manifold.
- This manifold is known as as the Lorenz Manifold.
$\circ$
- The Lorenz Manifold helps us understand the Lorenz attractor .

The Lorenz Equations: rho $=60$


Figure 69: Three orbits whose initial conditions agree to $>11$ decimal places !


Figure 70: A small portion of a Lorenz Manifold ...


Figure 71: Intersection of a Lorenz Manifold with a sphere ( $\rho=35, R=100$ ).


Figure 72: Intersection of a Lorenz Manifold with a sphere ( $\rho=35, R=100$ ).


Figure 73: Intersection of a Lorenz Manifold with a sphere ( $\rho=35, R=100$ ).

## How was the Lorenz Manifold computed?

First compute an orbit $\mathbf{u}_{0}(t)$, for $t$ from 0 to $T_{0}$ (where $T_{0}<0$ ), with
$\mathbf{u}_{0}(0)$ close to the origin $\mathbf{0}$,
and

$$
\mathbf{u}_{0}(0) \text { in the stable eigenspace spanned by } \mathbf{v}_{1} \text { and } \mathbf{v}_{2},
$$

that is,

$$
\mathbf{u}_{0}(0)=\mathbf{0}+\epsilon\left(\frac{\cos (2 \pi \theta)}{\left|\mu_{1}\right|} \mathbf{v}_{1}-\frac{\sin (2 \pi \theta)}{\left|\mu_{2}\right|} \mathbf{v}_{2}\right)
$$

for, say, $\theta=0$.

Scale time

$$
t \rightarrow \frac{t}{T_{0}}
$$

Then the initial orbit satisfies

$$
\mathbf{u}_{0}^{\prime}(t)=T_{0} \mathbf{f}\left(\mathbf{u}_{0}(t)\right), \quad \text { for } \quad 0 \leq t \leq 1,
$$

and

$$
\mathbf{u}_{0}(0)=\frac{\epsilon}{\left|\mu_{1}\right|} \mathbf{v}_{1} .
$$

The initial orbit has length

$$
L=T_{0} \int_{0}^{1}\left\|\mathbf{f}\left(\mathbf{u}_{0}(s)\right)\right\| d s
$$

Thus the initial orbit corresponds to a solution $\mathbf{X}_{0}$ of the equation

$$
\mathbf{F}(\mathbf{X})=\mathbf{0},
$$

where

$$
\mathbf{F}(\mathbf{X}) \equiv\left\{\begin{array}{l}
\mathbf{u}^{\prime}(t)-T \mathbf{f}(\mathbf{u}(t)) \\
\mathbf{u}(0)-\epsilon\left(\frac{\cos (\theta)}{\left|\mu_{1}\right|} \mathbf{v}_{1}-\frac{\sin (\theta)}{\left|\mu_{2}\right|} \mathbf{v}_{2}\right) \\
T \int_{0}^{1}\|\mathbf{f}(\mathbf{u})\| d s-L
\end{array}\right.
$$

with

$$
\mathbf{X}=(\mathbf{u}(\cdot), \theta, T), \quad(\text { for given } L \text { and } \epsilon),
$$

and

$$
\mathbf{X}_{0}=\left(\mathbf{u}_{0}(\cdot), 0, T_{0}\right)
$$

As before, the continuation system is

$$
\begin{aligned}
\mathbf{F}\left(\mathbf{X}_{k}\right) & =0 \\
<\mathbf{X}_{k}-\mathbf{X}_{k-1}, \dot{\mathbf{X}}_{k-1}>-\Delta s & =0, \quad\left(\left\|\dot{\mathbf{X}}_{k-1}\right\|=1\right)
\end{aligned}
$$

and

$$
\mathbf{X}=(\mathbf{u}(\cdot), \theta, T), \quad(\text { keeping } L \text { and } \epsilon \text { fixed })
$$

or

$$
\mathbf{X}=(\mathbf{u}(\cdot), \theta, L), \quad(\text { keeping } T \text { and } \epsilon \text { fixed }),
$$

or ( for computing the starting orbit $\mathbf{u}_{0}(t)$ ),

$$
\mathbf{X}=(\mathbf{u}(\cdot), L, T), \quad(\text { keeping } \theta \text { and } \epsilon \text { fixed }) .
$$

Other variations are possible ... .

NOTE:

- We do not just change the initial point (i.e., $\theta$ ) and integrate !
- Every continuation step requires solving a "boundary value problem".
- The continuation stepsize $\Delta s$ controls the change in $\mathbf{X}$.
- $\mathbf{X}$ cannot suddenly change a lot in any continuation step.
- This allows the "entire manifold" to be computed.

NOTE:

- Crossings of the Lorenz manifold with the plane $z=\rho-1$ can be located.
- Connections between the origin and the nonzero equilibria can be located.
- There are subtle variations on the algorithm !


Figure 74: Crossings of the Lorenz Manifold with the plane $z=\rho-1$ 282

## Heteroclinic connections

- During the computation of the 2D stable manifold of the origin one can locate heteroclinic orbits between the origin and the nonzero equilibria.
- The same heteroclinic orbits can be detected during the computation of the 2D unstable manifold of the nonzero equilibria.


Representation of the orbit family in the stable manifold.


A heteroclinic connection in the Lorenz equations. 285


Another heteroclinic connection in the Lorenz equations.

... and another ...
287

... and another ...
288


This continuation located 512 connections!

NOTE:

- The heteroclinic connections have a combinatorial structure.
- We can also continue each heteroclinic connection as $\rho$ varies.
- They spawns homoclinic orbits, having their own combinatorial structure.
- These results shed some light on the Lorenz attractor as $\rho$ changes.

More details:
E. J. Doedel, B. Krauskopf, H. M. Osinga, Global bifurcations of the Lorenz model, Nonlinearity 19, 2006, 2947-2972.

## Example: The Circular Restricted 3-Body Problem

$$
\begin{aligned}
x^{\prime \prime} & =2 y^{\prime}+x-\frac{(1-\mu)(x+\mu)}{r_{1}^{3}}-\frac{\mu(x-1+\mu)}{r_{2}^{3}} \\
y^{\prime \prime} & =-2 x^{\prime}+y-\frac{(1-\mu) y}{r_{1}^{3}}-\frac{\mu y}{r_{2}^{3}} \\
z^{\prime \prime} & =-\frac{(1-\mu) z}{r_{1}^{3}}-\frac{\mu z}{r_{2}^{3}}
\end{aligned}
$$

where

$$
(x, y, z)
$$

is the position of the zero-mass body, and

$$
r_{1}=\sqrt{(x+\mu)^{2}+y^{2}+z^{2}}, \quad r_{2}=\sqrt{(x-1+\mu)^{2}+y^{2}+z^{2}}
$$

For the Earth-Moon system $\mu \approx 0.01215$.

The CR3BP has one integral of motion, namely, the Jacobi-constant:

$$
J=\frac{\left(x^{\prime}\right)^{2}+\left(y^{\prime}\right)^{2}+\left(z^{\prime}\right)^{2}}{2}-U(x, y, z)-\mu \frac{1-\mu}{2},
$$

where

$$
U=\frac{1}{2}\left(x^{2}+y^{2}\right)+\frac{1-\mu}{r_{1}}+\frac{\mu}{r_{2}}
$$

and

$$
r_{1}=\sqrt{(x+\mu)^{2}+y^{2}+z^{2}}, \quad r_{2}=\sqrt{(x-1+\mu)^{2}+y^{2}+z^{2}}
$$

## BOUNDARY VALUE FORMULATION:

$$
\begin{aligned}
x^{\prime} & =T v_{x}, \\
y^{\prime} & =T v_{y}, \\
z^{\prime} & =T v_{z}, \\
v_{x}^{\prime} & =T\left[2 v_{y}+x-(1-\mu)(x+\mu) r_{1}^{-3}-\mu(x-1+\mu) r_{2}^{-3}+\lambda v_{x}\right], \\
v_{y}^{\prime} & =T\left[-2 v_{x}+y-(1-\mu) y r_{1}^{-3}-\mu y r_{2}^{-3}+\lambda v_{y}\right], \\
v_{z}^{\prime} & =T\left[-(1-\mu) z r_{1}^{-3}-\mu z r_{2}^{-3}+\lambda v_{z}\right],
\end{aligned}
$$

with periodicity boundary conditions

$$
\begin{gathered}
x(1)=x(0) \quad, \quad y(1)=y(0) \quad, \quad z(1)=z(0) \\
v_{x}(1)=v_{x}(0) \quad, \quad v_{y}(1)=v_{y}(0) \quad, \quad v_{z}(1)=v_{z}(0)
\end{gathered}
$$

+ phase constraint + continuation equation.

NOTE:

- The "unfolding term" $\lambda \nabla v$ regularizes the continuation.
- $\lambda$ will be "zero", once solved for.
- Other unfolding terms are possible.
- The unfolding term allows using BVP continuation.


Families of Periodic Solutions of the Earth-Moon system.


A family of Halo orbits.

NOTE:

- "Small" Halo orbits have one real Floquet multiplier outside the unit circle.
- Such Halo orbits are unstable.
- They have a 2D unstable manifold.


Continuation, keeping the endpoint $x$ (1) fixed.

NOTE:

- The unstable manifold can be computed by continuation .
- First compute a starting orbit in the manifold.
- Then continue the orbit keeping, for example, $x(1)$ fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x(1)$ fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed. 308


NEW1:Continuation, keeping the endpoint $x(1)$ fixed.


NEW1:Continuation, keeping the endpoint $x(1)$ fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


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NEW1:Continuation, keeping the endpoint $x(1)$ fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x(1)$ fixed.


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NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW1:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x(1)$ fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x(1)$ fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.


NEW2:Continuation, keeping the endpoint $x$ (1) fixed.

NOTE:

- Continuation with $x(1)$ fixed can lead to a Halo-to-torus connection.
- This heteroclinic connection can be continued as a solution to

$$
\begin{gathered}
\mathbf{F}\left(\mathbf{X}_{k}\right)=\mathbf{0} \\
<\mathbf{X}_{k}-\mathbf{X}_{k-1}, \dot{\mathbf{X}}_{k-1}>-\Delta s=0
\end{gathered}
$$

where

$$
\mathbf{X}=(\text { Halo orbit }, \text { Floquet function , connecting orbit }) .
$$

## Traveling Waves

- One can also use continuation to compute traveling wave phenomena.
- We illustrate this for a particular model from Biology.


## Wave Phenomena in a Distributed System

- We want to find traveling waves in a parabolic PDE.
- The PDE has one space dimension.
- Traveling waves are periodic solutions of a "reduced" ODE system.
- Solitary waves correspond to homoclinic orbits in the reduced system.
- Moving fronts are heteroclinic orbits in the reduced system.
- Thus ODE continuation techniques can be used.


## An Enzyme Model

- We consider an enzyme catalyzed reaction involving two substrates.
- The reaction takes place inside a single compartment.
- A membrane separates the compartment from an outside reservoir.
- In the reservoir the substrates are kept at a constant level.
- Enzymes are embedded in the membrane.
- The enzymes activate the reaction.

A simple model of such a reaction is

$$
\begin{aligned}
s^{\prime}(t) & =\left(s_{0}-s\right)-\rho R(s, a) \\
a^{\prime}((t) & =\alpha\left(a_{0}-a\right)-\rho R(s, a)
\end{aligned}
$$

- $\quad s$ and $a$ denote the concentrations of two chemical species.
- The reaction takes place inside a compartment.
- An excess of concentration of $s$ inhibits the reaction.
- a always activates the reaction.
- The reaction rate is proportional to

$$
R(s, a)=\frac{a}{\kappa_{1}+a} \frac{s}{1+s+\kappa_{2} s^{2}} .
$$

- $\quad s_{0}$ and $a_{0}$ are the constant concentrations in the outside reservoir .
- The reaction is catalyzed by an enzyme.

This equation has been used to model a reaction with substrates Oxaloacetate and NADH, and catalyzed by
Malate Deshydrogenase .

In this case appropriate parameter values are:

$$
\kappa_{1}=3.4 \quad, \quad \kappa_{2}=0.023 \quad, \quad \alpha=0.2
$$

Thus there are three parameters left, namely,

First we also fix

$$
s_{0} \quad, \quad a_{0} \quad, \quad \rho .
$$

and we use

$$
\rho=210,
$$

$$
a_{0} \text { as bifurcation parameter, }
$$

for each of the values

$$
s_{0}=143.0 \quad, \quad s_{0}=144.5 \quad, \quad s_{0}=145.0
$$



Bifurcation diagram of the ODE for $s_{0}=143.0, \rho=210$. Red: Stationary states ; Blue: Periodic orbits. The periodic family connects two Hopf bifurcations . Solid: Stable ; Dashed: Unstable.


Bifurcation diagram of the ODE for $s_{0}=144.5, \rho=210$.
The periodic family ends in a saddle-node homoclinic orbit.


Bifurcation diagram of the ODE for $s_{0}=145.0, \rho=210$.
The periodic family ends in a saddle homoclinic orbit.


The saddle homoclinic orbit that terminates the periodic family.

$$
\left(s_{0}=145.0 ;\right. \text { The two stationary points are also indicated.) }
$$

NOTE:

- For $s_{0}=143$ a family of stable periodic orbits connects the Hopf points.
- For $s_{0}=144.5$ there is only one Hopf bifurcation.
- At the other end the family ends in a saddle node homoclinic orbit.
- For $s_{0}=145$ there is also only one Hopf bifurcation.
- At the other end the family ends in a saddle homoclinic orbit.
- The bifurcation diagrams are shown superimposed in the next Figure.


The superimposed bifurcation diagrams.


Loci of folds, Hopf bifurcations, and homoclinic orbits.

## NOTE:

- The preceding 2-parameter diagram shows loci of "singular points".
- (Loci of folds, Hopf bifurcations, and homoclinic orbits .)
- There is a cusp on the locus of folds in the 2-parameter diagram.
- The Hopf bifurcation locus terminates on the fold locus near the cusp.
- At this end point the Hopf bifurcation has infinite period.
- (The steady state Jacobian has a double zero eigenvalue there.)
- (The geometric multiplicity of this eigenvalue is 1.)
- This singular point is called a Takens-Bogdanov (TB) bifurcation.

NOTE: continued...

- The locus of homoclinic orbits also emanates from the TB point.
- Part of the locus of homoclinic orbits follows the fold locus.
- These homoclinic orbits are called saddle-node homoclinic orbits .
- The stationary point on these homoclinic orbits is a fold point.
- (Thus this stationary point has a zero eigenvalue .)
- Compare the 2-parameter diagram to the 1-parameter diagrams !


## The Enzyme Model with Diffusion

Now consider the $s-a$ system with diffusion, namely, the PDE

$$
\begin{aligned}
s_{t} & =s_{x x}-\lambda\left[\rho R(s, a)-\left(s_{0}-s\right)\right] \\
a_{t} & =\beta a_{x x}-\lambda\left[\rho R(s, a)-\alpha\left(a_{0}-a\right)\right]
\end{aligned}
$$

with, as before, $a_{0}$ as a free parameter, and

$$
\rho=210 \quad, \quad \kappa_{1}=3.4 \quad, \quad \kappa_{2}=0.023
$$

and

$$
s_{0}=145 \quad, \quad \beta=5 \quad, \quad \lambda=3
$$

Look for traveling waves:

$$
s(x, t)=s(x-c t) \quad, \quad a(x, t)=a(x-c t)
$$

This reduces the PDE to two coupled ODEs:

$$
\begin{aligned}
s^{\prime \prime} & =-c s^{\prime}+\lambda\left[\rho R(s, a)-\left(s_{0}-s\right)\right] \\
a^{\prime \prime} & =-\frac{c}{\beta} a^{\prime}+\frac{\lambda}{\beta}\left[\rho R(s, a)-\alpha\left(a_{0}-a\right)\right] .
\end{aligned}
$$

Rewrite the reduced system

$$
\begin{aligned}
s^{\prime \prime} & =-c s^{\prime}+\lambda\left[\rho R(s, a)-\left(s_{0}-s\right)\right] \\
a^{\prime \prime} & =-\frac{c}{\beta} a^{\prime}+\frac{\lambda}{\beta}\left[\rho R(s, a)-\alpha\left(a_{0}-a\right)\right]
\end{aligned}
$$

as a first order system

$$
\begin{aligned}
s^{\prime} & =u \\
u^{\prime} & =-c u+\lambda\left[\rho R(s, a)-\left(s_{0}-s\right)\right] \\
a^{\prime} & =v \\
v^{\prime} & =-\frac{c}{\beta} v+\frac{\lambda}{\beta}\left[\rho R(s, a)-\alpha\left(a_{0}-a\right)\right]
\end{aligned}
$$

The stationary states satisfy

$$
\begin{aligned}
& u=v=0 \\
& \rho R(s, a)-\left(s_{0}-s\right)=0 \\
& \rho R(s, a)-\alpha\left(a_{0}-a\right)=0
\end{aligned}
$$

## NOTE:

- The stationary states do not depend on the wave speed $c$.
- The stationary states are those of the system without diffusion.
- The Jacobian of the stationary states now depends on $c$.
- Thus the Hopf bifurcations need not be present.
- However, for large $c$ there must be a Hopf bifurcation.
- The Hopf bifurcation approaches the ODE Hopf as $c$ gets large.
- Thus there are PDE wave trains for large $c$, when $s_{0}=145$. .
- The ODE homoclinic orbit implies PDE solitary waves for large $c$.


## Wave Trains and Solitary Waves

For the first bifurcation diagram for the reduced system we use $c=100$.

Indeed, we find that

- The stationary states are those of the system without diffusion.
- There is a Hopf bifurcation near the ODE Hopf bifurcation.
- The family of periodic orbits indeed ends in a homoclinic orbit.


Bifurcation diagram of the reduced system for $c=100$.

From the diagram for $c=100$ we can conclude that

- The PDE has wave trains of wave speed $c=100$.
- The PDE has a solitary wave of wave speed $c=100$.

Note that

- Stabilities are different from those for the system without diffusion.
- (The diagrams do not show stability now.)

Next:

- Are there low speed wave trains and low speed solitary waves ?
- To find out we compute the locus of homoclinics of the reduced system.
- As free parameters we use $a_{0}$ and the wave speed $c$.


The locus of solitary waves. (Shown for smaller values of the wave speed.)

From the preceding diagram we can draw the following conclusions :

- For wave speeds between 0.48 and 0.77 there are three solitary waves .
- These are at at different $a_{0}$-values, but have the same wave speed.
- Near $a_{0}=605$ there is a solitary wave of wave speed zero.
- (This is a stationary wave .)
- The circled special point will be discussed later.
- We first show a 1-parameter diagram for $c=0.4$.


Bifurcation diagram of the reduced system for $c=0.4$.
The blue branch represents wave trains.
Its homoclinic end point is a solitary wave.
Stability is not shown in this diagram.

For the diagram for $c=0.4$ we note that

- There is a fold on the wave train family.
- We can compute the locus of such folds for varying $a_{0}$ and $c$.
- We can also compute the locus of Hopf points for varying $a_{0}$ and $c$.
- We add these loci to the diagram with the locus of solitary waves .
- We also show 1-parameter diagrams for more values of $c$.


Homoclinic orbits (solitary waves), Hopf bifurcations, and folds.


Bifurcation diagram of the reduced system for $c=0.47$.


Bifurcation diagram of the reduced system for $c=0.5$.


Bifurcation diagram of the reduced system for $c=0.6$.


Bifurcation diagram of the reduced system for $c=0.7$.


Bifurcation diagram of the reduced system for $c=0.747$.

NOTE:

- There is a fold w.r.t. the wave speed $c$ on the solitary wave locus.
- (This fold is near $c=0.45, a_{0}=637.2$.)
- Two new solitary waves appear at that point.


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Now consider the circled point near $a_{0}=638.5, c=0.749$ :

- It then represents a transition from one solitary wave to another.
- In the diagram for $c=0.747$ note the near-vertical family.
- It is a wave-train family connecting the solitary waves 1 and 3 .
- At the section through the circled point the family is exactly vertical.



Three periodic solutions that represent solitary waves.
The independent variable has been scaled to the interval $[0,1]$.

## Traveling Waves on a Ring

- We locate traveling wave solutions of given wave length $L$.
- If necessary we put two or more waves in series to get wave length $L$.
- For the choice $L=22$ we found five distinct waves this way.
- We can continue these for varying $a_{0}$ and $c$, with $L=22$ fixed.
- This corresponds to computing traveling waves on a ring of size $L=22$.
- Results are shown projected onto the $a_{0}-c$-plane in the next diagram.


Loci of traveling wave solutions on a ring of size $L=22$.


A traveling wave on the ring. (The solution with label 7.)

## Stationary Waves on a Ring

- There are also families of stationary waves ("patterns") on the ring.
- Like traveling waves, these are not unique.
- (They can be phase shifted, i.e., rotated around the ring).
- Such patterns can be found by time-integrating unstable traveling waves.
- For example, traveling wave 11 is unstable.
- After time integration it approaches a stationary wave.
- This stationary wave is shown in the following diagram.


A stationary wave on the ring, obtained after time integration of the unstable traveling wave with label 11.

## NOTE:

- Stationary waves can be continued as traveling waves with $c=0$.
- A phase condition is necessary in this continuation.
- (Because stationary waves can be phase-shifted.)
- The next diagram shows a skeleton of the solution structure.


Traveling waves, stationary waves and uniform states

- The S-shaped curve with $c=0$ represents spatially uniform states.
- Other curves in the $c=0$ plane represent stationary waves.
- Curves that rise above the $c=0$ plane represent traveling waves.
- Time integration of unstable traveling waves gives other solutions .
- For example, waves bouncing off each other are found.
- Some of these are stable, while others are transient.

