

Continuation with respect to a variable

We have discussed methods for solving $f(x) = 0$, $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$.

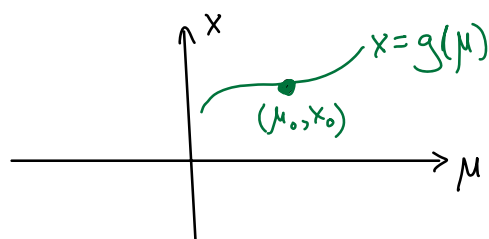
Now we consider one-parameter families

$$f(\mu, x), \quad f: \mathbb{R} \times \mathbb{R}^n \rightarrow \mathbb{R}^n.$$

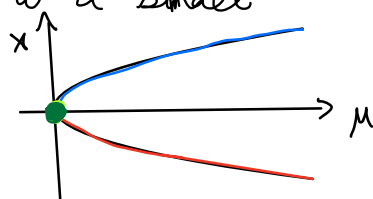
We wish to solve the parameter dependent equation $\{(\mu, x): f(\mu, x) = 0\}$.

Our goal is to discuss

continuation Given (μ_0, x_0) such that $f(\mu_0, x_0) = 0$
Finding nearby points (μ, x) with $f(\mu, x) = 0$.



bifurcation a qualitative change in dynamics occurring due to a small change in parameter



Application: $\dot{x} = f(\mu, x)$, $x(0) = x_0$

If $f(\mu_0, x_0) = 0$ then $x(t) \equiv x_0$ is an equilibrium solution for $\mu = \mu_0$.

Definition Hyperbolic matrix

a matrix $A \in \mathbb{R}^{n \times n}$ is hyperbolic if it has no eigenvalues with real part equal to zero.

Hyperbolic \Rightarrow non singular

Definition Topological equivalence

① $\dot{x} = f(x)$ $f: E_1 \rightarrow \mathbb{R}^n$ & ② $\dot{y} = g(y)$ $g: E_2 \rightarrow \mathbb{R}^n$

are called **topologically equivalent** if there is a function $H: E_1 \rightarrow E_2$ that preserves trajectories and their orientations.

That is, phase portraits of ① & ② are equivalent in terms of topological properties.

Definition

Bifurcation point for ODE

Consider

$$\dot{x} = f(\mu, x)$$

and equilibrium at (μ_0, x_0)

ie. $f(\mu_0, x_0) = 0$.

(μ_0, x_0) is a bifurcation point

there exist $\mu_k \rightarrow \mu_0$ such that if

$$\dot{x} = f(\mu_k, x)$$
 have phase portraits

which are not topologically equivalent
to $\dot{x} = f(\mu_0, x)$ near x_0 .

Theorem

Sufficient conditions for continuation

OR

Necessary conditions for a bifurcation

Let $f: \mathbb{R} \times \mathbb{R}^n \rightarrow \mathbb{R}^n$ be C^2 and assume

$\dot{x} = f(\mu, x)$ has an equilibrium at (μ_0, x_0) .

ie $f(\mu_0, x_0) = 0$

assume

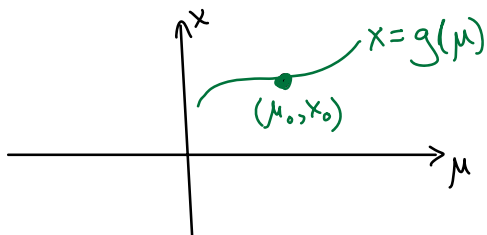
$D_x f(\mu_0, x_0)$ is hyperbolic

Then (μ_0, x_0) is not a bifurcation point.

Proof Since $D_x f(\mu_0, x_0)$ is nonsingular, the implicit function theorem implies that there exists $\delta > 0$ such that there is a function $g: [\mu_0 - \delta, \mu_0 + \delta] \rightarrow B_\delta(x_0)$ such that for all $\mu \in [\mu_0 - \delta, \mu_0 + \delta]$,

① $f(\mu, g(\mu)) = 0$

② this is the unique point in $B_\delta(x_0)$ with this property.



If we only care about existence of a unique path, we can stop here.

Top Conj: Since $D_x f(\mu_0, x_0)$ is hyperbolic, we can choose $\delta > 0$ so that

$D_x f(\mu, g(\mu))$ is hyperbolic for all $\mu \in B_\delta(\mu_0)$

Note that hyperbolic linear systems with the E^u of the same dimension are topologically conjugate. Thus by the Hartman-Grobman Theorem, there is no bifurcation in $B_\delta(\mu_0, x_0)$. \blacktriangle

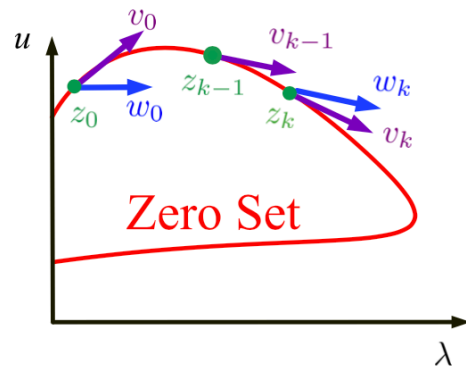
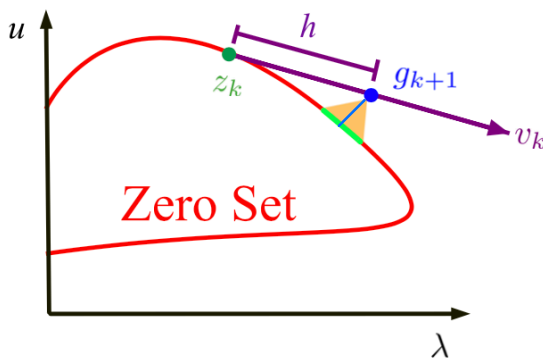
Assume ① $f(\mu_0, x_0) = 0$

② $D_x f(\mu_0, x_0)$ is non singular

How can we find a local part of $\{z = (\mu, x) : f(\mu, x) = 0\}$ given by $x = g(\mu)$

Predictor-corrector method

Numerical pseudo arclength continuation - iterative process



Predictor: Based on the previously known point on the zero set, find a prediction for the next point.

Corrector: Restrict to a line and use root finding to correct the original guess.

Pseudo arclength continuation

Predictor

Observe that $Z = \{z: F(z) = 0\}$
is orthogonal to $DF(z)$

\Rightarrow If $DF(z)v = 0$ then v is tangent to Z

Predictor g_{k+1} : Let z_k be on the zero curve

* Let h be small.

Define v_k to be a normalized tangent vector:

$$\begin{pmatrix} DF(z_k) \\ v_{k-1}^t \end{pmatrix} \tilde{v}_k = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix} \quad \&$$

$$v_k = \frac{\tilde{v}_k}{\|\tilde{v}_k\|}$$

Predict using the tangent line

$$g_{k+1} = z_k + h v_k$$

Corrector We make the assumption that z_{k+1} lies along the hypersurface orthogonal to the tangent line, solving the following **extended system**

$$G(z) = \begin{pmatrix} F(z) \\ v_k \cdot (z - g_{k+1}) \end{pmatrix} = 0 \quad G: \mathbb{R}^{(n+1)} \rightarrow \mathbb{R}^{(n+1)}$$

For Newton's method we also require DG

$$DG(z) = \begin{pmatrix} DF(z) \\ v_k^t \end{pmatrix} \in \mathbb{R}^{(n+1) \times (n+1)}$$

Code & simple examples as follows.

So far, this is the **non-validated** version. The validated version involves finding constructive conditions on the region for which a unique solution exists.

Matlab Code 5.5. *continuation.m*

```

1  % continuation (@filename,lambda,g,h)
2  % lambda is the first value of lambda
3  % g is the initial guess
4  % h is the stepsize
5  % The sign of h denotes the direction of the initial search
6
7  function z = continuation(fval,lambda,g,h)
8
9      tol = 0.0001;
10     MaxSteps = 150;
11
12     N = length(g)+1;
13     rvec = [zeros(N-1,1); 1];
14
15     % Find the first point by fixing lambda and using root finding
16     u = newton(fval,lambda,g,tol);
17     z(:,1) = [lambda;u];
18     v = sign(h)*[1;zeros(N-1,1)];
19
20     % Now the continuation loop using the extended system
21     for i = 2:MaxSteps
22         [y,yp,ylambda] = fval(z(:,i-1));
23         v = [ylambda yp; v'] \ rvec;
24         v = v/norm(v);
25         g = z(:,i-1) + abs(h)*v; % Continue in the direction of v, so use |h|
26         z(:,i) = newtonextended(fval,g,tol,g,v);
27     end
28
29     function u = newton(fval,lambda,guess,tolerance)
30         u = guess;
31         err = 1;
32         maxcount = 200;
33         count = 1;
34         while (err > tolerance & count < maxcount)
35             [y,yp] = fval([lambda;u]);
36             res = yp \ y;
37             err = max(abs(res));
38             u = u - res;
39             count = count + 1;
40         end
41
42     function z = newtonextended(fval,zguess,tolerance,g,v)
43         z = zguess;
44         err = 1;
45         maxcount = 200;
46         count = 1;
47         while (err > tolerance & count < maxcount)
48             [y,yp,ylambda] = fval(z);
49             yext = [y; dot(z-g,v)];
50             ypext = [ylambda, yp; v'];
51             res = ypext \ yext;
52             err = max(abs(res));
53             z = z - res;
54             count = count + 1;
55         end

```

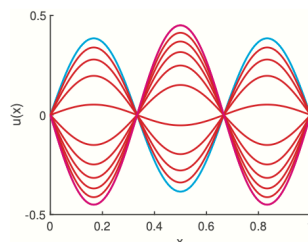
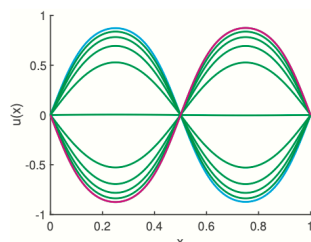
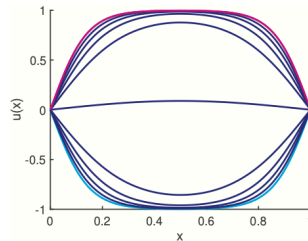
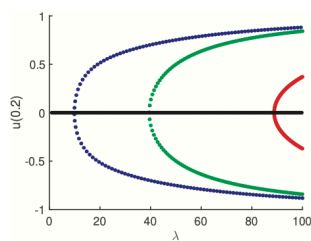
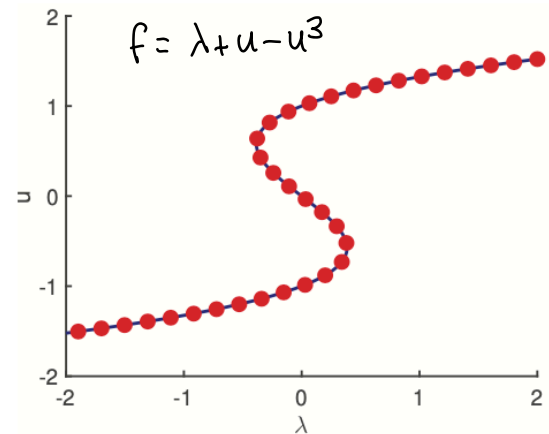
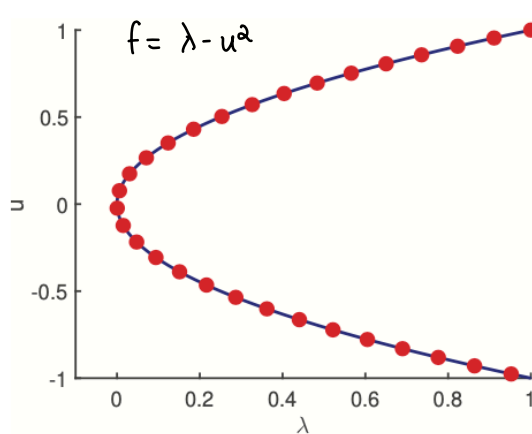
```

function [y,yp,ylambda] = quadratic(z)
lambda = z(1);
u = z(2);
y = lambda - u^2;
yp = -2*u; % derivative with respect to u
ylambda = 1; % derivative with respect to lambda

```

This implementation provides the function value $f(\lambda, u) = \lambda - u^2$, as well as the two partial derivatives $f_\lambda(\lambda, u) = 1$ with respect to λ and $f_u(\lambda, u) = -2u$ with respect to u . Once this file is accessible, the command

```
bif = continuation(@quadratic,1,1,-0.1);
```



$F(\lambda, u) = \Delta u + \lambda(u - u^3)$
using finite differences

Validated continuation

The proof of continuation relies on the Implicit Function Theorem

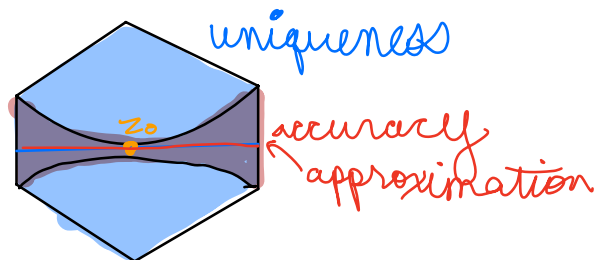
We now state a constructive version of this theorem that allows for validated continuation.

Theorem Constructive Implicit Function Theorem

assumptions

- ① $\|f(z_0)\| \leq \rho$ small residual $z = (\lambda, u)$
- ② assume $\|(DG(0,0))^{-1}\| \leq K$ (replacing ^{assuming} nonsingular)
- ③ assume Df is Lipschitz with constants
 $\|D_u f(\lambda, u) - D_u f(\lambda_0, u_0)\| \leq L_1 \|u - u_0\| + L_2 \|\lambda - \lambda_0\|$
 $\|D_\lambda f(\lambda, u)\| \leq L_3 + L_4 \|\lambda - \lambda_0\|$

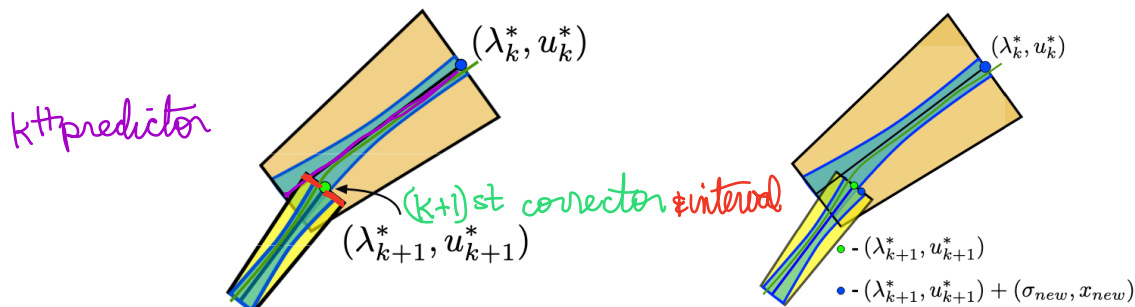
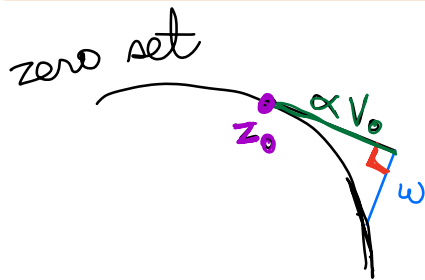
Then as long as $4K^2\rho < 1$, we have a list of specific conditions on existence and uniqueness of the root.



What about pseudo arclength continuation?

Use the theorem on the extended system

$$G(\alpha, \omega) = \begin{pmatrix} f(z_0 + \alpha v_0 + \omega) \\ v_0 \cdot \omega \end{pmatrix} = 0 \quad G: \mathbb{R}^{n+1} \rightarrow \mathbb{R}^{n+1}$$



Linking condition

Since we want to bound an entire path, we need to use many boxes. Thus we require $\alpha=0$ accuracy region of the $(k+1)$ st corrector point to be inside the k th uniqueness box.

another approach: series methods.
 write $u(\lambda)$ as a series and use root finding techniques.

Example $f(\lambda, u) = \lambda - u^2 = 0$ near $(1, 1)$

We seek to solve for $u(\lambda)$.

More conveniently we rescale and consider

$$1 + \lambda - (u(\lambda))^2 = 0 \quad \text{near } (0, 1)$$

Let $u(\lambda) = \sum_{j=0}^{\infty} a_j \lambda^j$ a Taylor series
 (where we see that $a_0 = 1$)

$$1 + \lambda - \sum_{j=0}^{\infty} (a * a)_j \lambda^j = 0$$

$$F(a) = 1 - a_0 + (1 - 2a_0 a_1) \lambda + \sum_{j=2}^{\infty} (a * a)_j \lambda^j = 0$$

$$F(a)_j = \begin{cases} 1 - a_0 & j = 0 \\ 1 - 2a_0 a_1 & j = 1 \\ -(a * a)_j = -\sum_{k=0}^j a_{j-k} a_k & j \geq 2 \end{cases}$$

Note that $(a * a)_j$ depends on a_0, \dots, a_j .

Therefore F_j depends only on a_0, \dots, a_j .

Consider the projection

$$F^N(a_0, \dots, a_N) = (F_0, \dots, F_N) \quad F^N: \mathbb{R}^{N+1} \rightarrow \mathbb{R}^{N+1}$$

If we find a_0, \dots, a_N such that $F^N(a_0, \dots, a_N) = 0$

How accurate is $u_N = \sum_{j=0}^N a_j \lambda^j$

Let $-\delta < \lambda < \delta$. Then $\|u - u_N\| \leq \sum_{j=N}^{\infty} |a_j| \delta^j$

We need to use a Newton-Kantorovich theorem on the operator

$$T(a) = a - A F(a)$$

where A is an approximate inverse of DF .

$$DF(a) = \begin{pmatrix} 1 & & & \\ 2a_1 & 2a_0 & & \\ 2a_2 & 2a_1 & 2a_0 & \\ 2a_3 & 2a_2 & 2a_1 & 2a_0 \\ \vdots & & & \ddots \end{pmatrix} = \begin{pmatrix} B_N & 0 \\ B_{N+1} & B_{N+2} \end{pmatrix}$$

Lower triangular

Define $A_N = B_N^{-1}$. Recall $a_0 = 1$.

Let A be given by $\begin{pmatrix} A_N & \\ & \ddots \end{pmatrix}$

Use techniques discussed to find the error estimates.

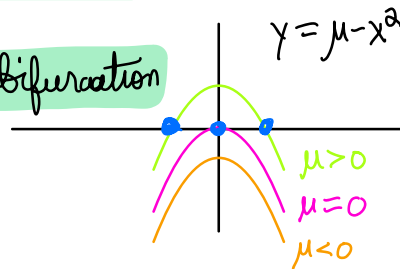
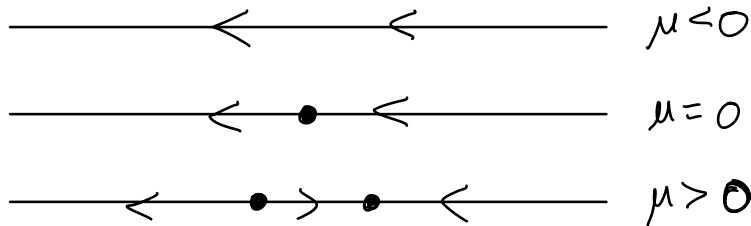
These notes will only occur if time allows.

Bifurcations of equilibria in 1D

Example Saddle-node bifurcation

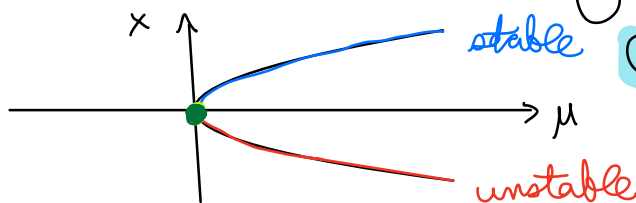
$$\dot{x} = \mu - x^2$$

Phase portraits



$$f(\mu, x) = \mu - x^2$$

Bifurcation diagram showing equilibria and their stability



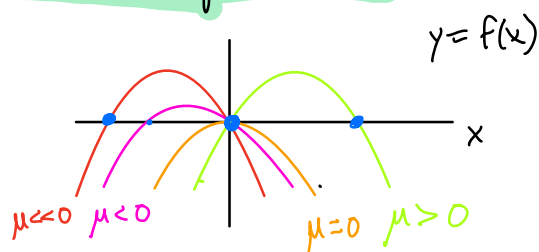
$(0,0)$ is a bifurcation point

Since the number of equilibria changes

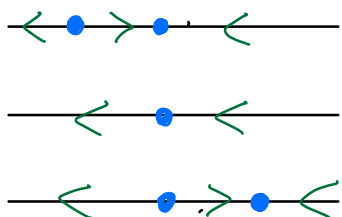
Example Transcritical bifurcation

$$\dot{x} = \mu x - x^2$$

$$f(\mu, x) = \mu x - x^2$$



Phase Portraits



$$f(0,0) = 0$$

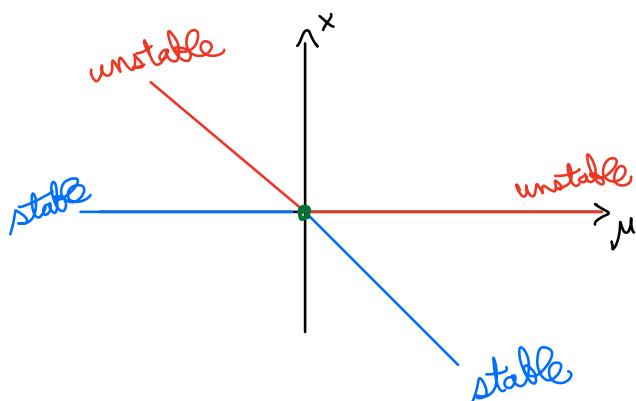
$$D_x f(0,0) = 0$$

$$D_\mu f(0,0) = 0$$

$$D_{x\mu} f(0,0) = 1 \neq 0$$

$$D_{xx} f(0,0) = -2 \neq 0$$

Bifurcation diagram



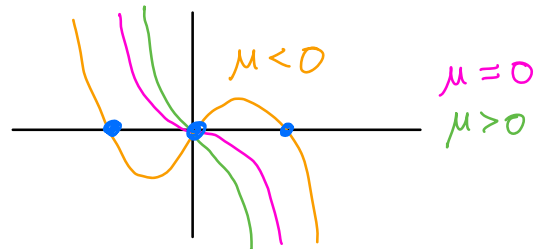
Exchange of stability
at $(0,0)$.

Example

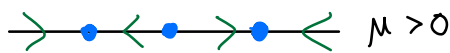
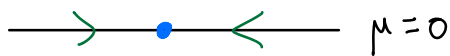
Pitchfork bifurcation

$$\dot{x} = \mu x - x^3$$

$$f(\mu, x) = \mu x - x^3$$



Phase Portraits



$$f(0,0) = 0$$

$$D_x f(0,0) = 0$$

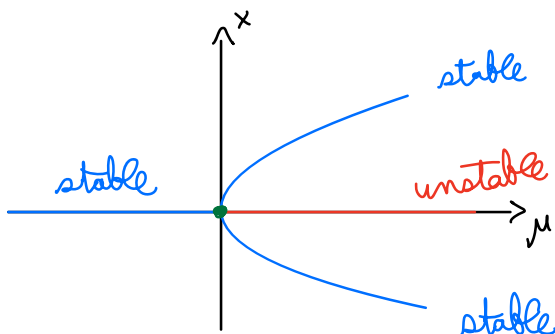
$$D_\mu f(0,0) = 0$$

$$D_{x\mu} f(0,0) = 1 \neq 0$$

$$D_{xx} f(0,0) = -6x|_{x=0} = 0$$

$$D_{xxx} f(0,0) = -6 < 0$$

Bifurcation diagram



Change of number of equilibria at $(0,0)$.

Calculus in higher dimensions

Define $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$

$D_x f$ is an $n \times n$ matrix

$D_x f(x_0) \cdot v \in \mathbb{R}^n$ directional derivative

How about higher order derivatives?

$$D_{xx} f(x_0)(u, v) = \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j} u_i v_j \in \mathbb{R}^n$$

$$D_{xxx} f(x_0)(u, v, w) = \sum_{i,j,k} \frac{\partial^3 f}{\partial x_i \partial x_j \partial x_k} u_i v_j w_k \in \mathbb{R}^n$$

Example

$$f(x_1, x_2) = \begin{pmatrix} x_1 x_2 + 7x_1^2 \\ x_2 \sin x_1 \end{pmatrix}$$

$$D_x f = \begin{pmatrix} x_2 + 14x_1 & x_1 \\ x_2 \cos x_1 & \sin x_1 \end{pmatrix}$$

$$\frac{\partial f}{\partial x_1^2} = \begin{pmatrix} 14 \\ -x_2 \sin x_1 \end{pmatrix}$$

$$\frac{\partial f}{\partial x_2^2} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\frac{\partial^2 f}{\partial x_1 \partial x_2} = \begin{pmatrix} 1 \\ \cos x_1 \end{pmatrix}$$

$$\begin{aligned} D_{xx} f(0,1) \left(\begin{pmatrix} 1 \\ 3 \end{pmatrix}, \begin{pmatrix} -2 \\ 4 \end{pmatrix} \right) &= \begin{pmatrix} 14 \\ 0 \end{pmatrix} 1(-2) + \begin{pmatrix} 0 \\ 0 \end{pmatrix} (3 \cdot 4) + \begin{pmatrix} 1 \\ 1 \end{pmatrix} (1 \cdot 4 + 3(-2)) \\ &= \begin{pmatrix} -28 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \end{pmatrix} + \begin{pmatrix} -2 \\ -2 \end{pmatrix} = \begin{pmatrix} -30 \\ -2 \end{pmatrix} \end{aligned}$$

Theorem Sufficient conditions for a bifurcation

Let $f: \mathbb{R} \times \mathbb{R}^n \rightarrow \mathbb{R}^n$ be C^2 , and

consider $\dot{x} = f(\mu, x)$ assume

- Extended System (μ_0, x_0, v)
- ① $f(\mu_0, x_0) = 0$
 - ② Singularity $A = D_x f(\mu_0, x_0)$ has a simple eigenvalue of 0
 $D_x f(\mu_0, x_0) v = 0$

Notation: right eigenvector v is $Av = 0 \cdot v$
left eigenvector w is $A^t w = w^t A = 0 w^t$

Then the following holds

- ① If ① $Q_1 = w^t D_\mu f(\mu_0, x_0) \neq 0$ Transversality
- ② $Q_2 = w^t D_{xx} f(\mu_0, x_0)(v, v) \neq 0$ Nondegeneracy

Then there is a saddle-node bifurcation at (μ_0, x_0) . Birth of 2 equilibria from none.

The sign of $\frac{Q_2}{Q_1}$ determines the direction it opens

$$\frac{Q_2}{Q_1} > 0 \Rightarrow \text{) } \quad \frac{Q_2}{Q_1} < 0 \Rightarrow \text{ (}$$

- ⑥ If
- ① $\omega^t D_\mu f(\mu_0, x_0) = 0$ *extra condition!*
 - ② $\omega^t D_{\mu x} f(\mu_0, x_0) v \neq 0$ *Transversality*
 - ③ $\omega^t D_{xx} f(\mu_0, x_0)(v, v) \neq 0$ *Nondegeneracy*

then there is a *transcritical bifurcation*.
 at (μ_0, x_0) . *i.e. Crossing of equil branches & exchange of stability*

- ⑦ If
- ① $\omega^t D_\mu f(\mu_0, x_0) = 0$
 - ② $\omega^t D_{\mu x} f(\mu_0, x_0) v \neq 0$ *Transversality*
 - ③ $\omega^t D_{xx} f(\mu_0, x_0)(v, v) = 0$
 - ④ $\omega^t D_{xxx} f(\mu_0, x_0)(v, v, v) \neq 0$ *Nondegeneracy*

then there is a *pitchfork bifurcation*
 at (μ_0, x_0) . *Birth of 2 new equilibria $1 \rightarrow 3$ with change of stability of one.*

Note: There are situations not covered by this theorem. These are all the typical cases with a simple zero. Another typical case is for A to have purely imaginary eigenvalues.

Example

$$\dot{x} = \mu + x - \ln(1+x)$$

$$f(0,0) = 0$$

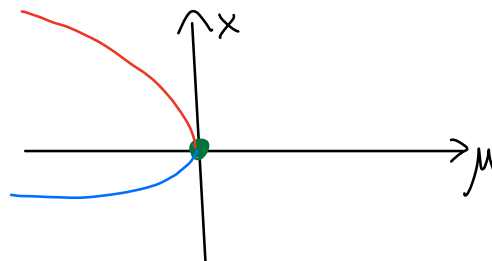
$$D_x f(0,0) = \left(1 + \frac{1}{1+x}\right)\bigg|_{x=0} = 0$$

$$D_\mu f(0,0) = 1 \neq 0$$

$$D_{xx} f(0,0) = +\frac{1}{(1+x)^2} = +1 \neq 0$$

$$\frac{D_{xx} f}{D_\mu f} = \frac{1}{1} > 0$$

\Rightarrow Saddle-node bifurcation at $(0,0)$
opening to the left.



Example

$$\dot{x} = \mu x - \ln(1+x)$$

$$f(1,0) = 0$$

$$D_x f(1,0) = \mu - \frac{1}{1+x} \Big|_{(1,0)} = 0$$

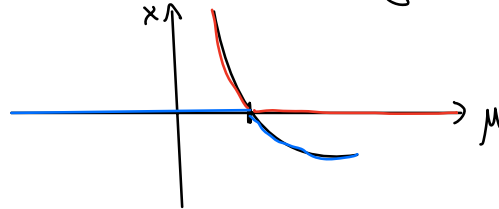
$$D_\mu f(1,0) = x \Big|_{(1,0)} = 0$$

$$D_{x\mu} f(1,0) = 1 \neq 0$$

$$D_{xx} f(1,0) = \frac{1}{(1+x)^2} \Big|_{(1,0)} = 1 \neq 0$$

Transcritical bifurcation

Bifurcation diagram



Example

$$\dot{x} = \mu x + x^3 - x^5$$

$$f(0,0) = 0$$

$$D_x f(0,0) = \mu + 3x^2 - 5x^4 \Big|_{(0,0)} = 0$$

$$D_\mu f(0,0) = x \Big|_{(0,0)} = 0$$

$$D_{\mu x} f(0,0) = 1 \neq 0$$

$$D_{xx} f(0,0) = 6x - 20x^3 \Big|_{(0,0)} = 0$$

$$D_{xxx} f(0,0) = 6 \neq 0$$

Pitchfork

$$\frac{f(\mu, x)}{x} = \mu + x^2 - x^4$$

the pitchfork opens rightward

