

Computation of bifurcation diagrams

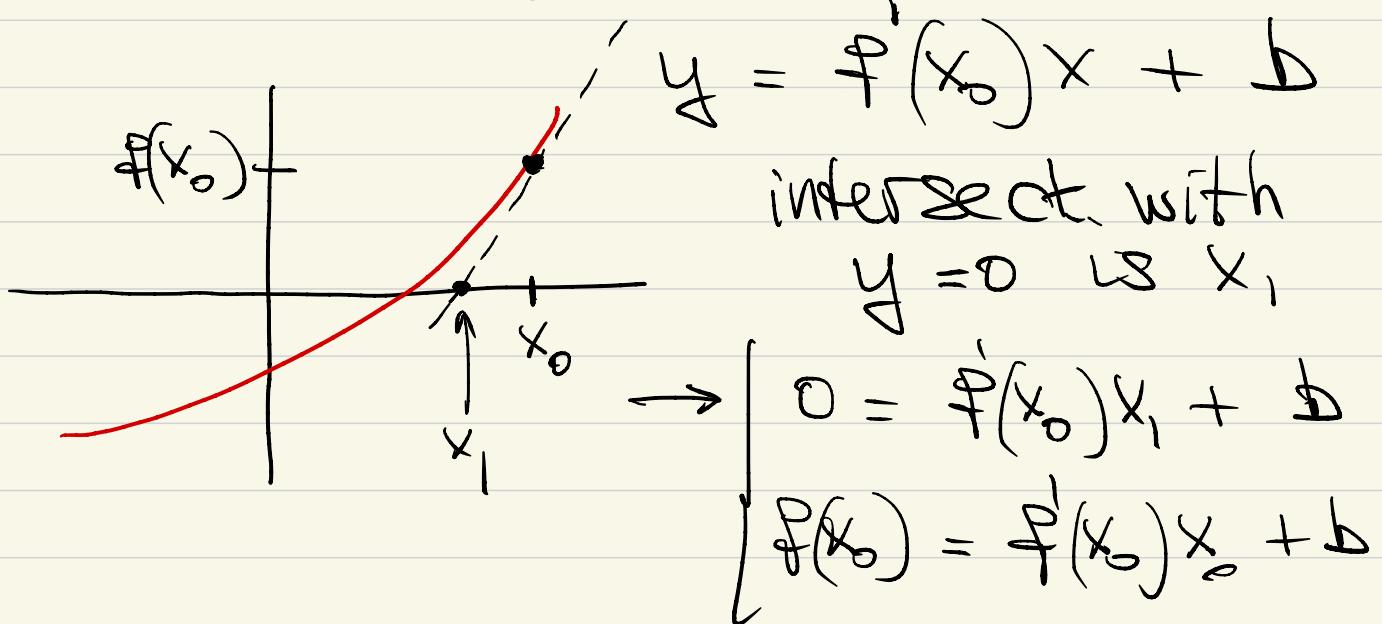
1. Fixed points

a) 1D : $f(\bar{x}, \lambda) = 0$
 zeroes of nonlinear eq.

Method: Newton - Raphson (NR)

iterate k

initial guess: x_0



$$\rightarrow f'(x_0)(x_1 - x_0) = -f(x_0)$$

assume $f'(x_0) \neq 0$

$$\rightarrow x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}$$

iteration k

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$$

NR converges quadratically
under 'mild' conditions to \bar{x} .

b) More generally:

$$\underline{f}(\bar{x}) = 0$$

$$\underline{f}(x_k) \left(\underline{x}_{k+1} - \underline{x}_k \right) = - \underline{f}(\underline{x}_k)$$

$$\Delta \underline{x}_{k+1}$$

\rightarrow system of linear equations

$$(A \underline{x} = \underline{b} + \text{trg})$$

Stability of fixed points \bar{x}

$$\underline{x} = \underline{x}^1 + \underline{x}^2, \quad \dot{\underline{x}} = F(\underline{x}, \lambda)$$

$$\rightarrow \dot{\underline{x}^2} = J(\bar{x}) \underline{x}^2 \quad (\text{page } 25)$$

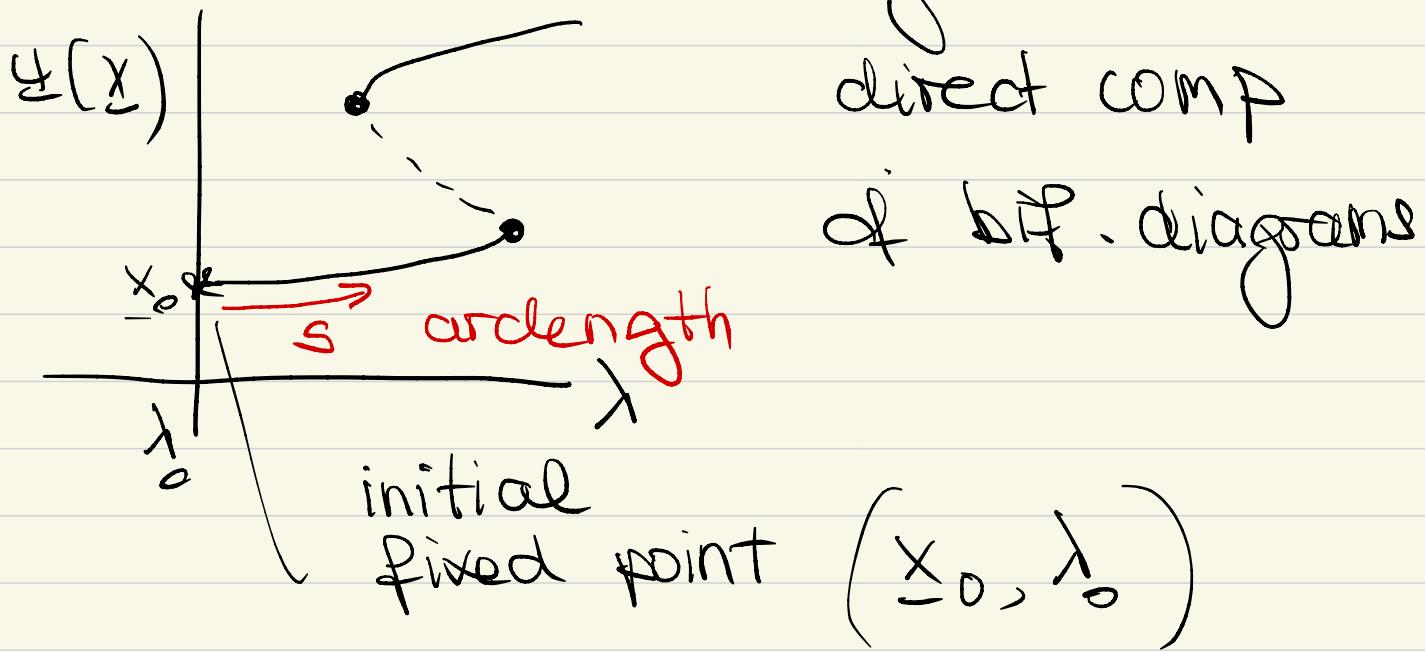
Let $\underline{x}^2 = e^{\sigma t} \underline{x}$

$$\rightarrow \sigma e^{\sigma t} \underline{x} = J(\bar{x}) e^{\sigma t} \underline{x}$$

$$\rightarrow J(\bar{x}) \underline{x} = \sigma \underline{x}$$

eigenvalue problem for
the matrix $J(\bar{x})$.

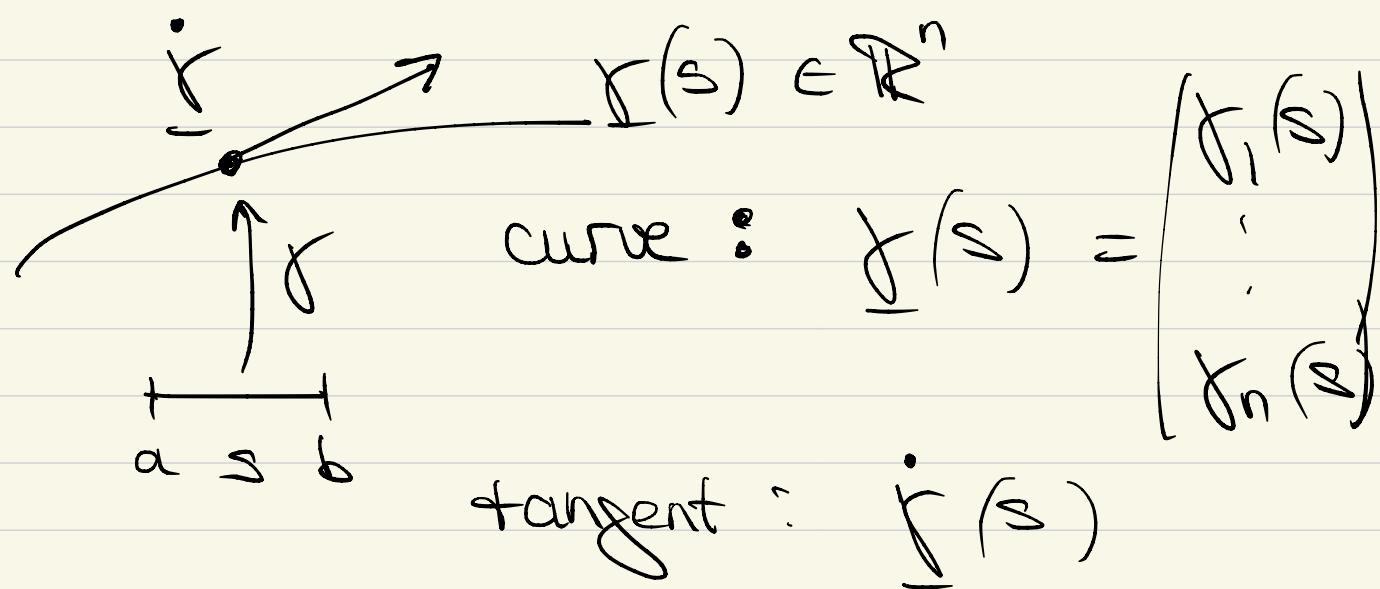
2. Pseudo - arclength methods



a) - Solutions: $(\underline{x}(s), \lambda(s))$
 $\text{of } \underline{\psi}(\underline{x}(s), \lambda(s)) = 0$

- Extra degree of freedom s

need extra condition Σ



$$\frac{\text{Ex}}{n=2} \quad \underline{x}(s) = \begin{pmatrix} \cos 2\pi s \\ \sin 2\pi s \end{pmatrix} \quad s \in [0, 1]$$

→ circle □

$$\underline{r}(s) = \begin{pmatrix} \underline{x}(s) \\ \lambda(s) \end{pmatrix}$$

$$\equiv (\underline{x}(s), \lambda(s)) = \left\| \underline{x}'(s) \right\|^2 - 1 = 0$$

→ normalization of the tangent.

Total system of equations:

$$f(\underline{r}(s)) = f(\underline{x}(s), \lambda(s)) = 0$$

$$\dot{\underline{x}}^T \dot{\underline{x}} + \lambda^2 = 1$$

b) Computation initial tangent

$$\dot{x}(s) = \begin{pmatrix} \dot{x}_0 \\ \dot{\lambda}_0 \end{pmatrix}$$

$$\text{Differentiate } \nabla f(x(s)) = 0$$

to get:

$$n \begin{bmatrix} \nabla & f_x \end{bmatrix} \dot{x}(s) = 0$$

$n \times n \quad n \times 1 \quad (n+1) \times 1$

- $\nabla f(x_0, \lambda_0)$ is not a df.

point then $\text{rank} [\nabla f_x] = n$

- We can determine $(\dot{x}_0, \dot{\lambda}_0)$ from

$$(n+1) \begin{pmatrix} \nabla(x_0) & f_x \\ 0 & I \end{pmatrix} \begin{pmatrix} \dot{x}_0 \\ \dot{\lambda}_0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

With steplength Δs , Σ is often approximated by

$$\Sigma = \dot{\underline{x}}_0^T \frac{\underline{x} - \underline{x}_0}{\Delta s} + \dot{\lambda}_0 \frac{\lambda - \lambda_0}{\Delta s} - 1$$

where (\underline{x}, λ) is the next fixed point to be computed. The system for $(\underline{x}(s), \lambda(s))$ is

$$\left\{ \begin{array}{l} \dot{\underline{x}}(s) = 0 \\ \dot{\lambda}(s) = \Delta s \end{array} \right.$$

$$\dot{\underline{x}}_0^T (\underline{x} - \underline{x}_0) + \dot{\lambda}_0 (\lambda - \lambda_0) = \Delta s$$

where Δs is chosen

and $\underline{x}_0, \dot{\underline{x}}_0, \lambda_0, \dot{\lambda}_0$
are known

c) Euler - Newton method

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- initial guess

$$\underline{x}_1 = \underline{x}_0 + \Delta s \dot{\underline{x}}_0$$

$$\lambda_1 = \lambda_0 + \Delta s \dot{\lambda}_0$$

- NR on (*)

$$\underline{x}_{k+1} = \underline{x}_k + \Delta \underline{x}_{k+1}$$

$$\begin{pmatrix} I \\ \underline{x}_0^T \end{pmatrix} \begin{pmatrix} \mathcal{F} \\ \dot{\lambda} \end{pmatrix} \begin{pmatrix} \Delta \underline{x}_{k+1} \\ \Delta \lambda_{k+1} \end{pmatrix} = \begin{pmatrix} I \\ \alpha \end{pmatrix}$$

$$\begin{pmatrix} I \\ \alpha \end{pmatrix} = \begin{pmatrix} -\mathcal{F} & (\underline{x}_k, \lambda_k) \\ \Delta s - \underline{x}_0^T & (\underline{x}_k - \underline{x}_0) - \lambda_0(1-\delta) \end{pmatrix}$$

Advantage : at bif points

J becomes singular but \mathcal{F} does not.

Converged solution gives next

fixed point $(x(s_1), \lambda(s_1))$

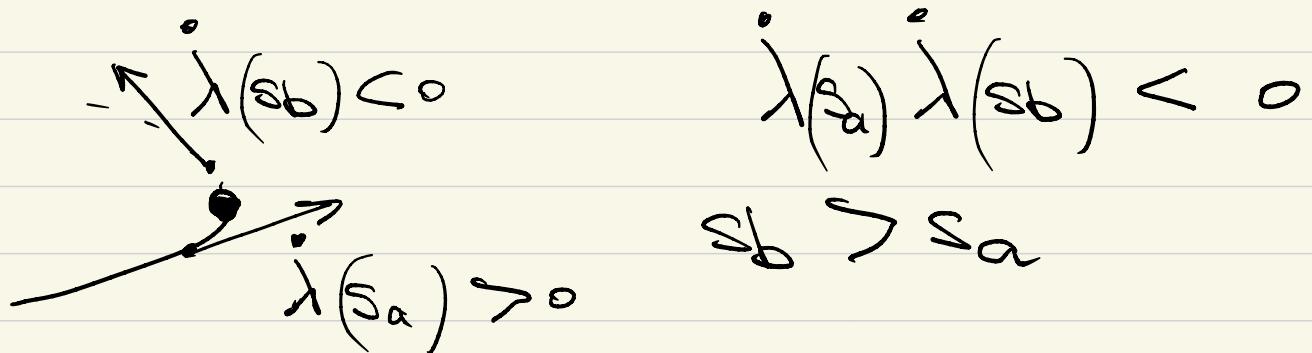
with $s_1 = s_0 + \Delta s$, when

$$\left\| \begin{pmatrix} \Delta x_{k+1} \\ \Delta \lambda_{k+1} \end{pmatrix} \right\| < \epsilon_{NR} \left\| \begin{pmatrix} x_0 \\ \lambda_0 \end{pmatrix} \right\|$$

- NR determines fixed points regardless of their stability

3. Detection of bifurcation points

- saddle-node:



- pitchfork | transcritical

$$1) \left(\det \bar{J}(s_a) \right) \cdot \left(\det \bar{J}(s_b) \right) < 0$$

2) solve stability problem

$$\bar{J} \left(\underline{x}(s), \lambda(s) \right) \dot{\underline{x}} = \bar{J} \dot{\underline{x}}$$

$$\bar{\sigma} = \bar{\sigma}_r \pm i \bar{\sigma}_i$$

then bif when: $\bar{\sigma}_r(s_a), \bar{\sigma}_r(s_b) < 0$

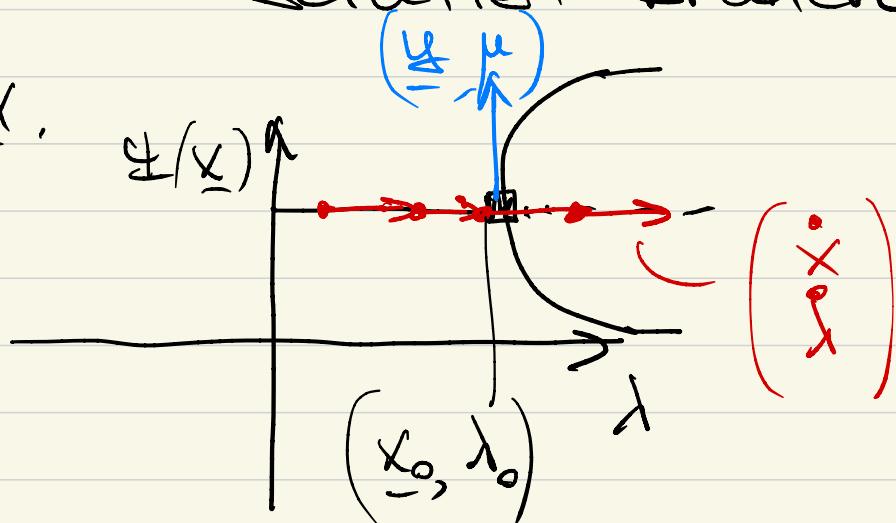
After a sign change in any indicator $\phi(s)$ has been found, the precise location of the bif point can be computed through a secant iteration:

$$\left\{ \begin{array}{l} s_{e+1} = s_e - \frac{s_e - s_{e-1}}{\phi(s_e) - \phi(s_{e-1})} \\ s_0 = s_a, \quad s_1 = s_b \end{array} \right.$$

until $|s_{e+1} - s_e| < \epsilon_s s_0$

4. Detection of additional solution branches.

EK.



- Branch switching near BP

solution (x_0, λ_0)

$$\text{compute } \begin{pmatrix} y \\ \mu \end{pmatrix} \perp \begin{pmatrix} x_0 \\ \lambda_0 \end{pmatrix}$$

and $\nabla V = 0$

$$\rightarrow \begin{pmatrix} \nabla & f_0 \\ x_0^T & \lambda_0^T \end{pmatrix} \begin{pmatrix} y \\ \mu \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$(\ast\ast) \rightarrow \begin{cases} \nabla y + \mu f_0 = 0 \\ x_0^T y + \mu \lambda_0 = 0 \end{cases}$$

Let $\underline{\boldsymbol{\xi}}$ be the solution of

$$\mathcal{J} \underline{\boldsymbol{\xi}} = \underline{\mathbf{f}}$$

which can be solved as the kernel

$\underline{\mathbf{V}}$ is known



$$\left\{ \begin{array}{l} \underline{\mathbf{y}} = \underline{\mathbf{V}} - \mu \underline{\boldsymbol{\xi}} \\ \underline{\mathbf{x}}_0^T \underline{\mathbf{V}} - \mu \underline{\mathbf{x}}_0^T \underline{\boldsymbol{\xi}} + \mu \lambda_0 = 0 \end{array} \right.$$

$$\rightarrow \mu = \frac{-\underline{\mathbf{x}}_0^T \underline{\mathbf{V}}}{\lambda_0 - \underline{\mathbf{x}}_0^T \underline{\boldsymbol{\xi}}}$$

Start Euler - Newton with

$$\left\{ \begin{array}{l} \underline{\mathbf{x}}_1 = \underline{\mathbf{x}}_0 + \Delta s \underline{\mathbf{y}}_1 \\ \lambda_1 = \lambda_0 + \Delta s \mu \end{array} \right.$$

In practise, other methods are often used.

5. Stability of branches

- zelle $\exists \left(x(s), \lambda(s) \right) \dot{x} = \sigma \dot{x}$

standard OR method.