

# Notes on Solving Non-Linear Equation Systems in OpenModelica

## Status and Plans on Solving Non-Linear Equation Systems

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*You've never heard of Chaos theory? Non-linear equations? Strange attractors?*

*Michael Crichton, Jurassic Park*

- Current status:
  - ▶ Which methods do we use?
  - ▶ How are details solved?
- Work in progress

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- 1 Systems of Non-Linear Equations
- 2 What are the crucial points?
  - Initial Values
  - Derivatives
- 3 Current Status in OpenModelica
  - Newton Solver
  - Hybrid Solver
  - Homotopy Solver

### Transformation steps for simulation

$$\underline{0} = \underline{f}(\underline{x}(t), \underline{\dot{x}}(t), \underline{y}(t), \underline{u}(t), \underline{p}, t)$$

↓

$$\underline{0} = \underline{f}(\underline{x}(t), \underline{z}(t), \underline{u}(t), \underline{p}, t), \underline{z}(t) = \begin{pmatrix} \underline{\dot{x}}(t) \\ \underline{y}(t) \end{pmatrix})$$

↓

$$\underline{z}(t) = \begin{pmatrix} \underline{\dot{x}}(t) \\ \underline{y}(t) \end{pmatrix} = \underline{g}(\underline{x}(t), \underline{u}(t), \underline{p}, t)$$

↓

$$\underline{\dot{x}}(t) = \underline{h}(\underline{x}(t), \underline{u}(t), \underline{p}, t)$$

$$\underline{y}(t) = \underline{k}(\underline{x}(t), \underline{u}(t), \underline{p}, t)$$

### Transformation example

$$f_2(z_2) = 0$$

$$f_4(z_1, z_2) = 0$$

$$f_3(z_2, z_3, z_5) = 0$$

$$f_5(z_1, z_3, z_5) = 0$$

$$f_1(z_3, z_4) = 0$$

### Algebraic loop (SCC)

$$f_3(z_2, z_3, z_5) = 0$$

$$f_5(z_1, z_3, z_5) = 0$$

## General form

$$f_1(x_1, x_2, \dots, x_n) = 0$$

$$f_2(x_1, x_2, \dots, x_n) = 0$$

⋮

$$f_m(x_1, x_2, \dots, x_n) = 0$$

in compact matrix form:

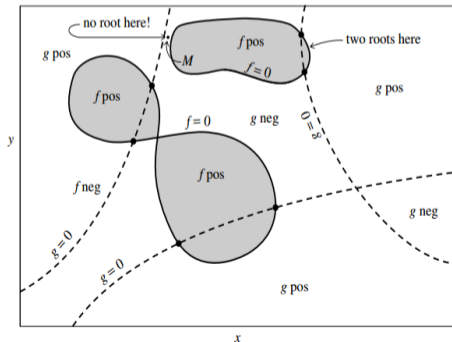
$$\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$$

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

## Example

$$f_1(x_1, x_2) = 0$$

$$f_2(x_1, x_2) = 0$$

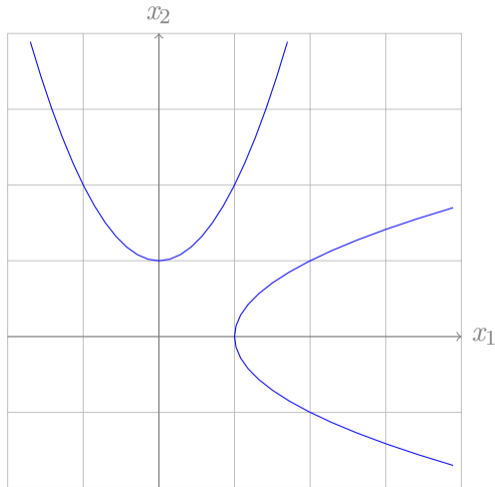


### Example

$$\begin{aligned}x_1^2 - x_2 + \alpha &= 0 \\ -x_1 + x_2^2 + \alpha &= 0\end{aligned}$$

### Solutions

- $\alpha = 1 \Rightarrow$  no solution.
- $\alpha = \frac{1}{4} \Rightarrow$  one solution  $x_1 = x_2 = \frac{1}{2}$ .
- $\alpha = 0 \Rightarrow$  two solutions  $x_1 = x_2 = 0$  and  $x_1 = x_2 = 1$ .
- $\alpha = -1 \Rightarrow$  four solutions.

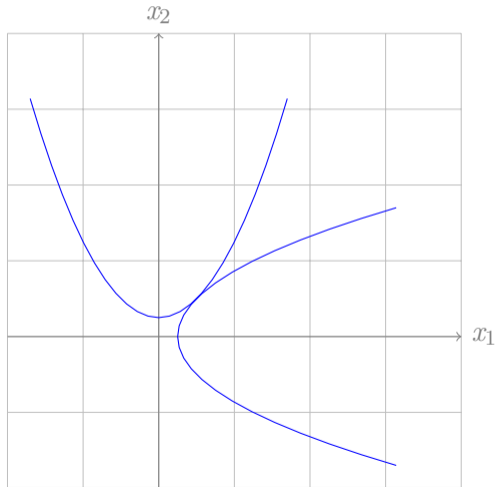


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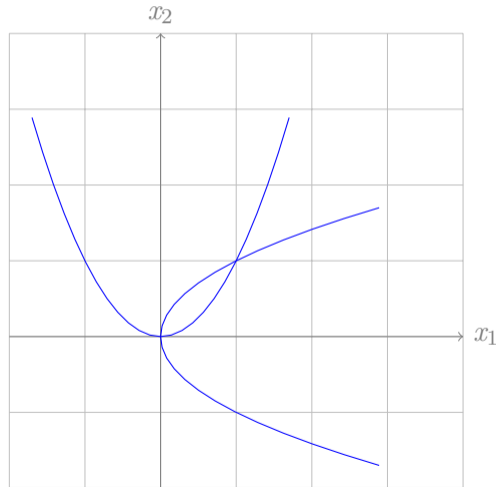


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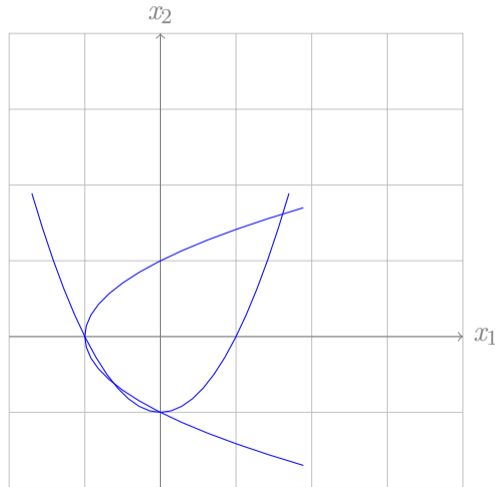


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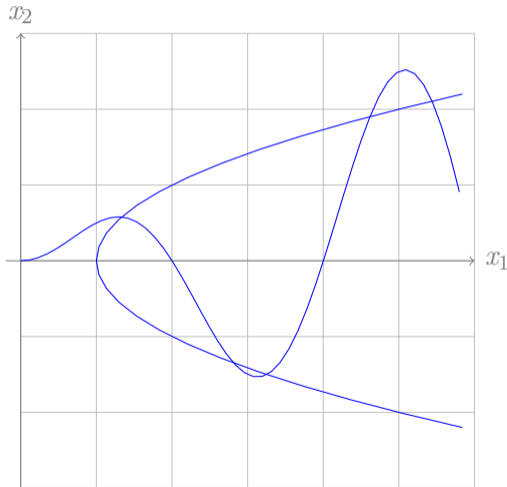


### Example

$$\begin{aligned}\frac{1}{2}x_1 \sin\left(\frac{1}{2}\pi x_1\right) - x_2 &= 0 \\ -x_1 + x_2^2 + 1 &= 0\end{aligned}$$

### solution

- Countable infinite many solutions.



# What are the crucial points?

Common things

## General form

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

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

Iteration instruction:

$$\underline{\mathbf{x}}_{i+1} = \Phi(\underline{\mathbf{x}}_i)$$

Initial Values:  $\underline{\mathbf{x}}_0 \in \mathbb{R}^n$

## Crucial points

- Good initial values
- Quite accurate derivatives
- Fast linear solver

## General problems

- Round-off effects and cancellation
- Modelica asserts

# What are the crucial points?

Initial Values

Model with an algebraic loop

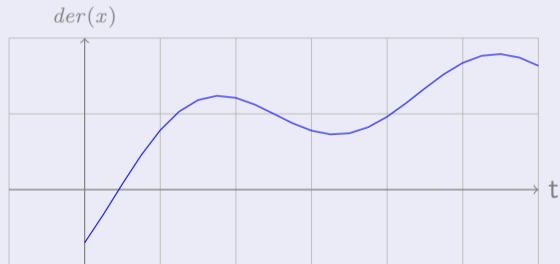
$$f_1(y_1, y_2) = 0$$

$$f_2(y_1, y_2) = 0$$

within

$$\dot{\underline{x}}(t) = \text{functionODE}(\underline{x}(t), \underline{u}(t), \underline{y}(t), \underline{p}, t)$$

## Trajectory



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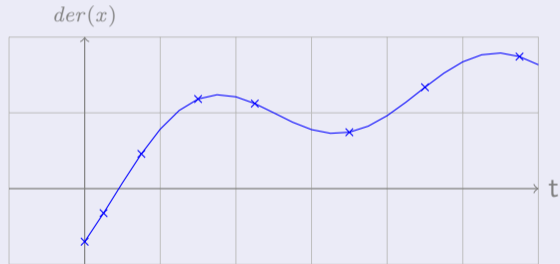
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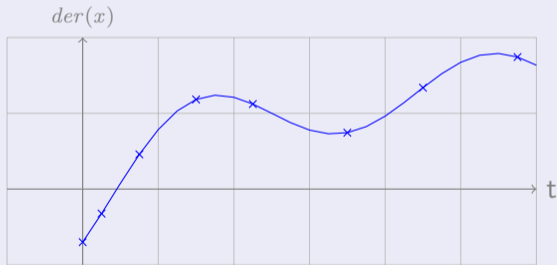
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## Extrapolation

$$y(t_*) = y_{k-1} + \frac{t_* - t_{k-1}}{t_k - t_{k-1}} (y_k - y_{k-1})$$

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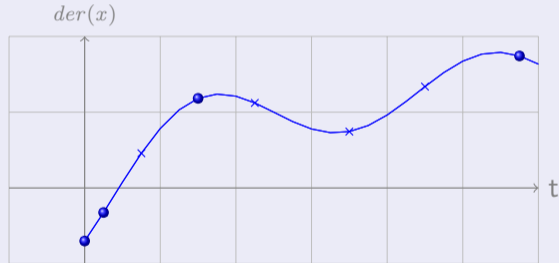
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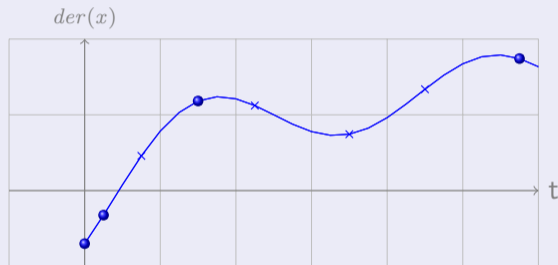
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## Jacobian evaluations



## Solution

Every non-linear loop uses now lists for **context dependent extrapolation**.

# What are the crucial points?

## Derivatives

$$\dot{f}(x) = \frac{f(x + \delta) - f(x)}{\delta}$$

### Drawback

Even if  $\delta$  is selected optimal:

$$\left| \frac{\partial f(x)}{\partial x} - \frac{f(x + \delta_{opt}) - f(x)}{\delta_{opt}} \right| \approx \sqrt{\epsilon_{RND}}$$

Some significant digits get lost by truncation.

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### Approach

Tried to sort such sums by nominal values, if states and derivatives are involved.

+heats=derCalls

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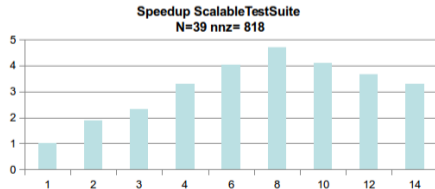
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⇒ Use of symbolic jacobians with parallelization -jacPar=N



### Damped Newton algorithm

- Damping strategy based on expected function decrease and validity of iteration step

### Newton iteration step:

$$\begin{aligned} \underline{J}_F(\underline{x}^{(k)}) \cdot \underline{s}^{(k)} &= -\underline{F}(\underline{x}^{(k)}) \\ \underline{x}^{(k+1)} &= \underline{x}^{(k)} + \tau^{(k)} \underline{s}^{(k)} \end{aligned}$$

Damping parameter:

$$\tau^{(k)} \in [0, 1].$$



### MinPack: HYBRID method

- Based on Powell's method, which is based on conjugate direction method.
- Through an Newton corrector step it has the same the convergence rate as it.
- Robustness through QR-Solver.
- $11.5O(n^2)$

### Possible homotopy functions

- **Fixpoint-Homotopy:**

$$\underline{H}(\underline{x}, \lambda) = \lambda \underline{F}(\underline{x}) + (1 - \lambda)(\underline{x} - \underline{x}_0) = \underline{0}$$

- **Newton-Homotopy:**

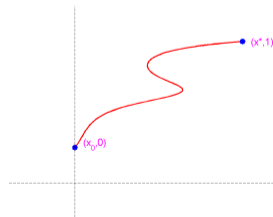
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### Simple example

$$f(x) = 2x - 4 + \sin(2\pi x),$$

$$x_0 = 0.5, \quad x^* = 2.$$

### Homotopy Path (Fixpoint)



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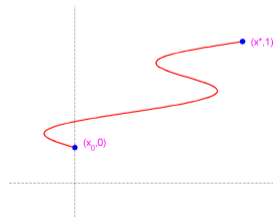
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### Homotopy Path (Newton)



### Newton

- homotopy
  - ▶ Current default.
  - ▶ Total pivot as linear solver
  - ▶ Combined damping strategy with Modelica Asserts
- kinsol
  - ▶ Not in our hand.
  - ▶ KLU as linear solver
  - ▶ Currently no combination with the homotopy solver.
- newton
  - ▶ Current testing ground.
  - ▶ Lapack as linear solver
  - ▶ Currently no combination with the homotopy solver.

### Hybrid

- Former default solver.
- Linpack as linear solver
- Developed initial value selection.
- Currently no combination with the homotopy solver.

- Finalize the equality of arms and compare!
- Includes same handling of initial values and homotopy combination.
- Compare own newton vs. kinsol with same linear solver.

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