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Solving Differential Equations under Uncertainty

Guaranteed Numerical Integration of Nonlinear Parametric ODEs

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In this talk, we present a Mathematica implementation of algorithms which compute reachable space for nonlinear dynamical systems in presence of model uncertainty. These algorithms will be described through an example of a biological process, discussing the quality of the results and the efficiency for getting them.

■ Non-Linear Dynamical Systems

□ Generic Case

$$\{\dot{x}(t) = f(x, p, t), \quad x(t_0) \in X_0 \subseteq D, \quad p \in P\} \quad (1)$$

□ Example

Haldane model of a biotechnological process in a stirred reactor, addressing the existence of one species on a chemostat with a single substrate

$$\begin{cases} \dot{x} = f_x(x, s) = \left(\mu_0 \frac{s}{s + k_s + s^2/k_i} - \alpha d \right) x \\ \dot{s} = f_s(x, s) = -k \mu_0 \frac{s}{s + k_s + s^2/k_i} x + (s_{in} - s) d \end{cases} \quad (2)$$

x : biomass density,

s : substrate concentration,

d : dilution rate of the chemostat,

s_{in} : concentration of input substrate given by

$s_{in}(t) = s_{in}^0 + 15 \cos(1/5 t)$.

μ_0 and s_{in}^0 are assumed uncertain.

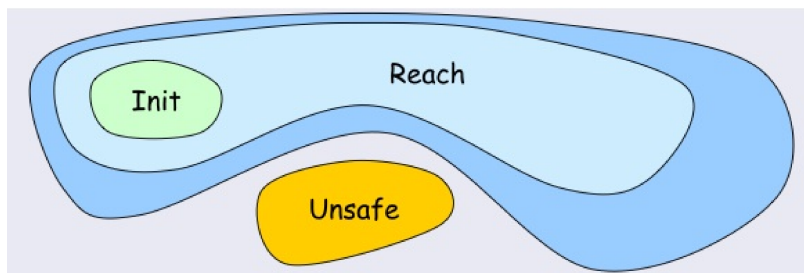
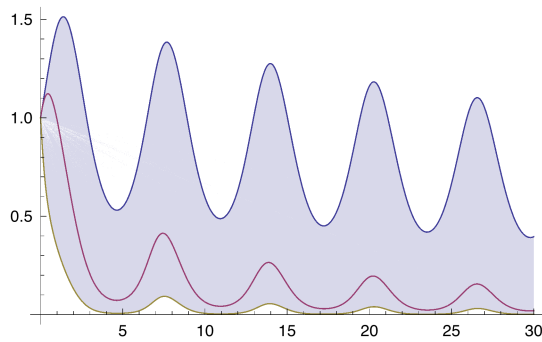
■ Reachable Set

$$\{\dot{x}(t) = f(x, p, t), x(t_0) \in X_0 \subseteq D, p \in P\} \tag{3}$$

The reachable space is the set of all feasible solutions for the differential equation when initial state vector is taken in domain X_0 and parameter vector in domain P

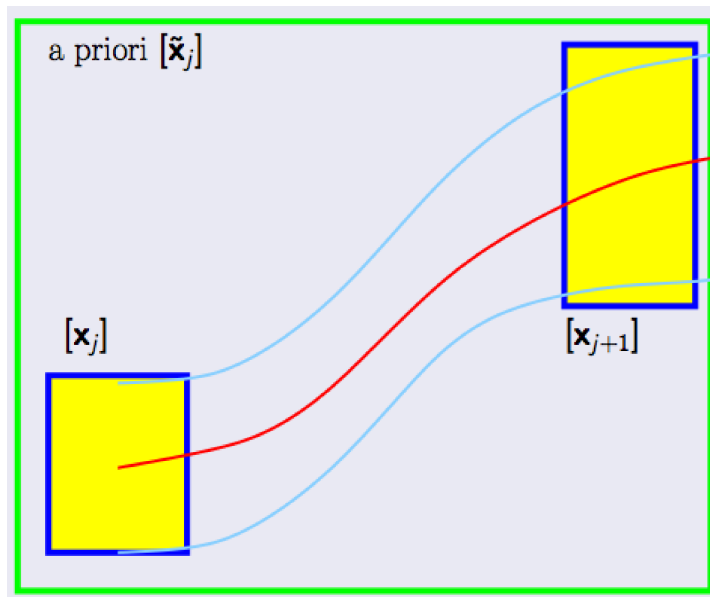
$$R([t_1, t_2]; X_0) = \{x(\tau), t_1 \leq \tau \leq t_2 \mid (\dot{x}(\tau) = f(x, p, \tau) \wedge x(t_0) \in X_0 \wedge p \in P)\} \tag{4}$$

```
s[α_] := NDSolve[
  {y'[x] == α y[x] Cos[x + α^ (3) y[x]], y[0] == 1}, y, {x, 0, 30}]
Plot[Evaluate[{y[x] /. s[0.5], y[x] /. s[1.], y[x] /. s[1.5]}],
  {x, 0, 30}, PlotRange -> All, Filling -> {1 -> {3}}]
```



■ Set Integration with Interval Taylor Models

- define a time grid $t_0 < t_1 < t_2 < \dots < t_{n_T}$
- objective is to compute interval vectors $[x_j]$, $j = 1, \dots, n_T$, that are *guaranteed* to contain the solution at time t_j



1. Fixed point theorem and Picard–Lindelöf operator, compute an *a priori* enclosure $[\tilde{x}_j]$

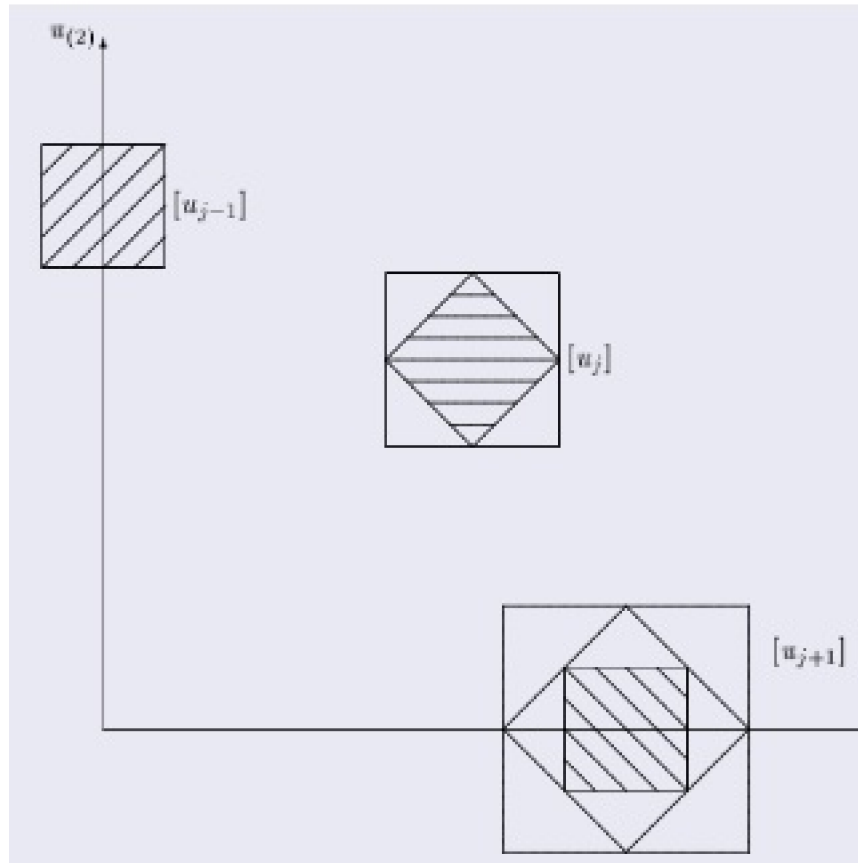
$$\forall t \in [t_j, t_{j+1}] \quad x(t) \in [\tilde{x}_j] \quad (5)$$

2. a tighter enclosure $[x_{j+1}]$ of the solution is based on Taylor expansion

$$[x_{j+1}] = [x_j] + \sum_{i=1}^{k-1} h_j^i f^{(i)}([x_j], [p], t_j) + h_j^k f^{(k)}([\tilde{x}_j], [p], [t_j, t_{j+1}]) \quad (6)$$

■ Set Integration with Interval Taylor Models (cont'd)

□ Wrapping Effect



□ Implementation

- a C++ package which implements Rhim's algorithm (mean value forms, matrices preconditioning and linear transforms)
- Profil/BIAS C++[17] class library for interval computations
- FADBAD++ package is used for computing Taylor coefficients

■ Set Integration with Müller's Theorem

□ Müller's Theorem

- Hypothesis :

$$\begin{aligned}\forall i, \quad D^{\pm} \omega_i(t) &\leq \min_{\underline{T}_i(t)} f_i(x, p, t) \\ \forall i, \quad D^{\pm} \Omega_i(t) &\geq \max_{\overline{T}_i(t)} f_i(x, p, t)\end{aligned}\tag{7}$$

$$\begin{aligned}\underline{T}_i &: \left\{ x_i = \omega_i(t), \quad \omega_j(t) \leq x_j \leq \Omega_j(t), \quad j \neq i, \quad \underline{p} \leq p \leq \overline{p} \right\} \\ \overline{T}_i &: \left\{ x_i = \Omega_i(t), \quad \omega_j(t) \leq x_j \leq \Omega_j(t), \quad j \neq i, \quad \underline{p} \leq p \leq \overline{p} \right\}\end{aligned}\tag{8}$$

- Solution of system is inside

$$X: \begin{cases} t_0 \leq t \leq t_{nr} \\ \omega(t) \leq x(t) \leq \Omega(t) \end{cases}\tag{9}$$

□ Using Monotonicity

- Study sign of the partial derivatives $\frac{\partial f_i}{\partial p_k}$ and $\frac{\partial f_i}{\partial x_j}$
- Solve

$$\begin{cases} \dot{\omega}(t) = \underline{f}(\omega, \Omega, \underline{p}, \overline{p}, t), \quad \omega(t_0) = \underline{x}_0 \\ \dot{\Omega}(t) = \overline{f}(\omega, \Omega, \underline{p}, \overline{p}, t), \quad \Omega(t_0) = \overline{x}_0 \end{cases}\tag{10}$$

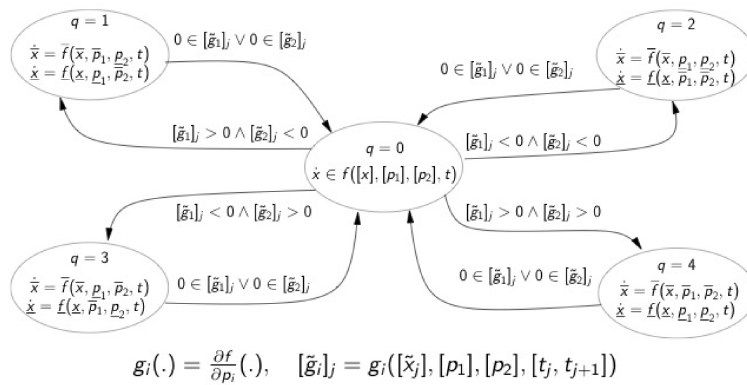
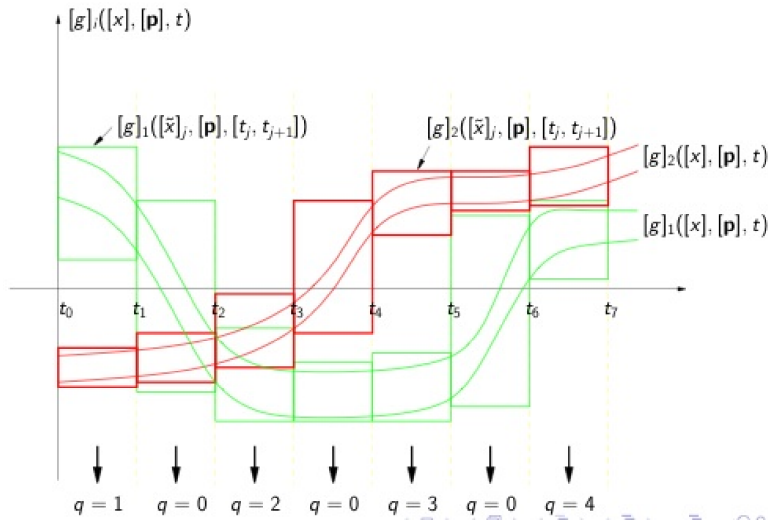
with

$$\begin{aligned}\underline{f}_i(\omega, \Omega, \underline{p}, \overline{p}, t) &= f_i(\underline{\gamma}^i(x), \underline{\delta}^i(p), t) \\ \overline{f}_i(\omega, \Omega, \underline{p}, \overline{p}, t) &= f_i(\overline{\gamma}^i(x), \overline{\delta}^i(p), t)\end{aligned}\tag{11}$$

$$\overline{\gamma}^i(x_j) = \begin{cases} \Omega_j & \text{if } i = j \\ \Omega_j & \text{if } (i \neq j) \wedge \frac{\partial f_i}{\partial x_j} \geq 0 \\ \omega_j & \text{if } (i \neq j) \wedge \frac{\partial f_i}{\partial x_j} < 0 \end{cases}\tag{12}$$

Hybridization

- Use Müller's theorem whenever it is possible, use interval Taylor models otherwise



■ Practical Case

```

Uncert[v_, p_] := Interval[N[{v - p v/100, v + p v/100}]]
si[t] := s0 + p1 Cos[t/5]
equations = { (mu0 s / (s + ks + s^2/ki) - alpha d) x,
              -k mu0 s / (s + ks + s^2/ki) x + (si[t] - s) d }
variables = {x -> Interval[{0.00000001, 1000}],
             s -> Interval[{0.00000001, 1000}]};
parameters = {mu0 -> Uncert[0.75, 1], s0 -> Uncert[65, 1.5]};
constants = {k -> 42.14, ks -> 9.28, ki -> 256, alpha -> 0.5, d -> 2., p1 -> 15};

```

$$\left\{ x \left(\frac{\mu_0 s}{ks + s + \frac{s^2}{ki}} - d \alpha \right), -\frac{k \mu_0 s x}{ks + s + \frac{s^2}{ki}} + d \left(-s + s_0 + p_1 \cos\left[\frac{t}{5}\right] \right) \right\}$$

□ Derivatives and signs

```

(jac = D[equations, {First /@ Join[variables, parameters]}]) //
MatrixForm

```

$$\begin{pmatrix} \frac{\mu_0 s}{ks + s + \frac{s^2}{ki}} - d \alpha & -\frac{\mu_0 s \left(1 + \frac{2s}{ki}\right)}{\left(ks + s + \frac{s^2}{ki}\right)^2} + \frac{\mu_0}{ks + s + \frac{s^2}{ki}} x & \frac{s x}{ks + s + \frac{s^2}{ki}} & 0 \\ -\frac{k \mu_0 s}{ks + s + \frac{s^2}{ki}} & -d + \frac{k \mu_0 s \left(1 + \frac{2s}{ki}\right) x}{\left(ks + s + \frac{s^2}{ki}\right)^2} - \frac{k \mu_0 x}{ks + s + \frac{s^2}{ki}} & -\frac{k s x}{ks + s + \frac{s^2}{ki}} & d \end{pmatrix}$$

```

(sigjac = Fold[ReplacePart[#1, 1, {#2, #2}] &,
              Map[Sign, jac /. Join[{t -> Interval[{0, 30}], constants} /.
              parameters /. variables, {2}],
              Range[Length[jac]]]) // MatrixForm

```

$$\begin{pmatrix} 1 & \text{Interval}[-1, 1] & 1 & 0 \\ -1 & 1 & -1 & 1 \end{pmatrix}$$

□ Code

Building equations for each mode

Given an equation, the corresponding line of the sign matrix, and uncertain variables, **BuildEquation** returns the corresponding bounding equations. Minima and maxima of variable x are denoted by new variables $xMin$ and $xMax$.


```

BuildEquation[eq_, l_, v_] :=
  eq /. Transpose[Map[{{#[[1]] → #[[2, 1]], #[[1]] → #[[2, 2]]} &,
    Inner[{{#2, SetMm[#1][#2]} &, l, v, List]]]
SetMm[1] = {min[#, max[#]} &;
SetMm[-1] = {max[#, min[#]} &;
SetMm[0] = {#, #} &;
min[s_Symbol] := ToExpression[ToString[s] <> "Min"]
max[s_Symbol] := ToExpression[ToString[s] <> "Max"]

```

For a given mode criterion, and the corresponding sign matrix, **BuildEquations** applies **BuildEquation** for generating all the equations of the mode.

```

BuildEquations[{mc_, sig_}, var_, param_, eqs_] :=
  {mc, MapIndexed[BuildEquation[#1, sig[[#2[[1]]]],
    First /@ Join[var, param]] &, eqs]}

```

Building modes

BuildMode is the low level function computing all the modes and the corresponding equations.

```

BuildMode[jac_, p_, sig_, var_, param_, eqs_] :=
  Map[BuildEquations[#, var, param, eqs] &,
    Transpose[{With[{l = Map[{{# > 0, # < 0}} &, Extract[jac, p]]},
      Flatten[Outer[List, Apply[Sequence, l]], Length[l] - 1]],
      Map[BuildSignJac[#, sig] &,
        With[{l = Map[{{toto[1, #], toto[-1, #]} &, p}}, Flatten[
          Outer[List, Apply[Sequence, l]], Length[l] - 1]]]]]}]
BuildSignJac[{x__toto}, sig_] :=
  Fold[ReplacePart[#1, #2[[1]], #2[[2]]] &, sig, {x}]

```

GenMode is the high level function computing all the modes and the corresponding equations.

```

GenMode[var_, param_, eq_, val_] :=
  With[{jac = D[eq, {First /@ Join[var, param]}]},
    With[{sigjac = Fold[ReplacePart[#1, 1, {#2, #2}] &, Map[Sign,
      jac /. val /. param /. var, {2}], Range[Length[jac]]]},
      With[{pos = Position[sigjac, _Interval]},
        BuildMode[jac, pos, sigjac, var, param, eq]]]}]

```

```

GenJacMode[var_, param_, eq_, val_] := Map[FullSimplify,
  With[{jac = D[eq, {First /@ Join[var, param]}]}],
  With[{sigjac = Fold[ReplacePart[#1, 1, {#2, #2}] &,
    Map[Sign, jac /. val /. param /. var, {2}],
    Range[Length[jac]]]}, With[
    {pos = Position[sigjac, _Interval]}, Extract[jac, pos]]]]]

```

Result is a list of mode. Each mode is a {<criteria>, <equations>} list. <criteria> is a list of inequations. <equations> is a list of couples of {min, max} equations.

□ Mode Computation

```

res = GenMode[variables, parameters,
  equations, Join[{t → Interval[{0, 30]}], constants]]

```

$$\left\{ \left\{ \left\{ -\frac{\mu_0 s \left(1 + \frac{2s}{k_i}\right) + \frac{\mu_0}{ks + s + \frac{s^2}{k_i}}}{\left(ks + s + \frac{s^2}{k_i}\right)^2} x > 0 \right\}, \right. \right.$$

$$\left. \left\{ x_{\text{Min}} \left(\frac{\mu_{0\text{Min}} s_{\text{Min}}}{ks + s_{\text{Min}} + \frac{s_{\text{Min}}^2}{k_i}} - d \alpha \right), x_{\text{Max}} \left(\frac{\mu_{0\text{Max}} s_{\text{Max}}}{ks + s_{\text{Max}} + \frac{s_{\text{Max}}^2}{k_i}} - d \alpha \right) \right\}, \right.$$

$$\left. \left\{ -\frac{k \mu_{0\text{Max}} s_{\text{Min}} x_{\text{Max}}}{ks + s_{\text{Min}} + \frac{s_{\text{Min}}^2}{k_i}} + d \left(s_{0\text{Min}} - s_{\text{Min}} + p_1 \cos\left[\frac{t}{5}\right] \right), \right. \right.$$

$$\left. \left. -\frac{k \mu_{0\text{Min}} s_{\text{Max}} x_{\text{Min}}}{ks + s_{\text{Max}} + \frac{s_{\text{Max}}^2}{k_i}} + d \left(s_{0\text{Max}} - s_{\text{Max}} + p_1 \cos\left[\frac{t}{5}\right] \right) \right\} \right\} \right\},$$

$$\left\{ \left\{ \left\{ -\frac{\mu_0 s \left(1 + \frac{2s}{k_i}\right) + \frac{\mu_0}{ks + s + \frac{s^2}{k_i}}}{\left(ks + s + \frac{s^2}{k_i}\right)^2} x < 0 \right\}, \right. \right.$$

$$\left. \left\{ x_{\text{Min}} \left(\frac{\mu_{0\text{Min}} s_{\text{Max}}}{ks + s_{\text{Max}} + \frac{s_{\text{Max}}^2}{k_i}} - d \alpha \right), x_{\text{Max}} \left(\frac{\mu_{0\text{Max}} s_{\text{Min}}}{ks + s_{\text{Min}} + \frac{s_{\text{Min}}^2}{k_i}} - d \alpha \right) \right\}, \right.$$

$$\left. \left\{ -\frac{k \mu_{0\text{Max}} s_{\text{Min}} x_{\text{Max}}}{ks + s_{\text{Min}} + \frac{s_{\text{Min}}^2}{k_i}} + d \left(s_{0\text{Min}} - s_{\text{Min}} + p_1 \cos\left[\frac{t}{5}\right] \right), \right. \right.$$

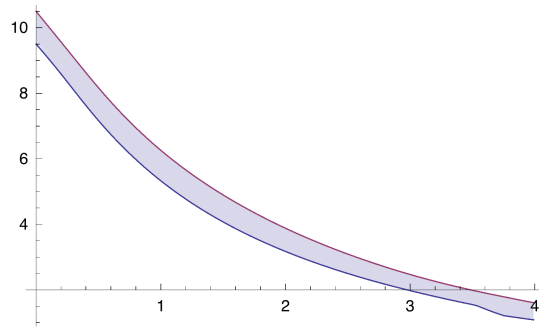
$$\left. \left. -\frac{k \mu_{0\text{Min}} s_{\text{Max}} x_{\text{Min}}}{ks + s_{\text{Max}} + \frac{s_{\text{Max}}^2}{k_i}} + d \left(s_{0\text{Max}} - s_{\text{Max}} + p_1 \cos\left[\frac{t}{5}\right] \right) \right\} \right\} \right\}$$

□ C code linking

Code generation

□ Results

PlotFrontier [1, 1]



PlotFrontier [1, 2]

