Nonlinear Parameter Continuation with COCO

Lecture given during Advanced Summer School on Continuation Methods for Nonlinear Problems

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Outline

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1 An Example of Functional Optimization





Consider the autonomous two-point boundary-value problem

$$\dot{x}_1 = x_2, \, \dot{x}_2 = (1 + x_2^2)/x_1, \, x_1(0) = 1, \, x_1(1) = p$$

in terms of the vector of state variables $x = (x_1, x_2) \in \mathbb{R}^2$ and the scalar problem parameter $p \in \mathbb{R}$. Solutions correspond to extremal curves $s \mapsto f(s)$, and their derivatives, for the integral functional

$$\int_0^1 f(s) \sqrt{1 + f'(s)^2} \, \mathrm{d}s$$

in the space of functions that satisfy the boundary conditions f(0) = 1 and f(1) = p.

Derive Euler-Lagrange's equations for the functional!

For arbitrary initial conditions $x_1(0)$ and $x_2(0)$, solutions to the associated initial-value problem are given by

$$x_1(t) = \frac{x_1(0)}{\sqrt{1 + x_2^2(0)}} \cosh\left(\frac{\sqrt{1 + x_2^2(0)}}{x_1(0)}t + \operatorname{arcsinh} x_2(0)\right)$$

and

$$x_2(t) = \sinh\left(\frac{\sqrt{1+x_2^2(0)}}{x_1(0)}t + \operatorname{arcsinh} x_2(0)
ight).$$

The shape is that of a *catenary curve*.

Verify by substitution!

Recall that $x_1(0) = 1$ and $x_1(1) = p$.

For each p, a solution to the boundary-value problem then corresponds to a solution of the nonlinear equation

$$\frac{1}{\sqrt{1+x_2^2(0)}} \cosh\left(\sqrt{1+x_2^2(0)} + \operatorname{arcsinh} x_2(0)\right) = \rho.$$

Since the left-hand side is convex with a unique global minimum at $x_2(0) \approx -2.26$, it follows that there are no solutions to the boundary-value problem for $p \lesssim 0.587$ and two solutions for $p \gtrsim 0.587$.

Graph left-hand side as a function of $x_2(0)!$

Alternatively, approximate x_1 and x_2 by two polynomials

$$p_1(t) := \sum_{k=0}^m a_k t^k, \ p_2(t) := \sum_{k=0}^m b_k t^k$$

where

$$p_1(0) = 1, \ p_1(1) = p$$

and

$$p_1'(t_i) = p_2(t_i), \, p_2'(t_i) = (1 + p_2^2(t_i))/p_1(t_i)$$

for some sequence of *collocation nodes* $\{t_i\}_{i=1}^m$ on the interval [0, 1].

Write out equations for m = 2!

Alternatively, approximate f by the polynomial

$$p(t) := \sum_{k=0}^m a_k t^k,$$

such that $abla F(a_0,\ldots,a_m,\lambda_1,\lambda_2)=0$, where

$$F(a_0, ..., a_m, \lambda_1, \lambda_2) := \sum_{i=1}^n w_i \left(p(t_i) \sqrt{1 + p'^2(t_i)} \right) - \lambda_1(p(0) - 1) - \lambda_2(p(1) - p)$$

for some sequences of quadrature weights $\{w_i\}_{i=1}^n$ and quadrature nodes $\{t_i\}_{i=1}^n$ on [0, 1].

Write out equations for m = 2!

The MATLAB-based Computational Continuation Core (COCO) enables an approximate analysis of the catenary boundary-value problem, even without access to the closed-form solution.

We construct a family of approximate solutions for admissible values of p by

- constructing a family of approximate trajectory segments that satisfy x₁(0) = 1 and x₂(0) = 0, but are defined only on the interval [0, T] for T ∈ [0, 1];
- constructing a family of approximate trajectory segments on the interval [0, 1] that satisfy x₁(0) = 1.

Here, continuation implements the classical method of shooting.

Encode the vector field in the anonymous function ${\tt cat},$ as shown in the following command

>> cat = @(x,p) [x(2,:); (1+x(2,:).^2)./x(1,:)];

The encoding is vectorized and autonomous.

A corresponding trajectory segment is given by the single-point time history assigned below to the t0 and x0 variables.

```
>> t0 = 0;
>> x0 = [1 0];
```

Here, t0 encodes a one-dimensional array of time instances and x0 encodes a two-dimensional array of the corresponding points in state space, with one row per time instant.

We compute a family of trajectory segments under variations in T by invoking the coco entry-point function as shown in the sequence of commands below.

```
>> prob = coco_prob();
>> prob = ode_isol2coll(prob, '', cat, t0, x0, []);
>> data = coco_get_func_data(prob, 'coll', 'data');
>> maps = data.coll_seg.maps;
>> prob = coco_add_pars(prob, 'pars', ...
        [ maps.x0_idx; maps.x1_idx(1); maps.T_idx ], ...
        { 'y1s' 'y2s' 'y1e' 'T' });
>> cont_args = { 1, { 'T' 'y1e' }, [0 1] };
>> coco(prob, 'coll1', [], cont_args{:});
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```

Suppose that the function

$$\Phi: \mathbb{R}^n \mapsto \mathbb{R}^m, \ n \ge m \ge 1$$

is continuously differentiable. The equation

 $\Phi(u)=0$

is a *(continuation) zero problem* in the vector u of unknown *continuation variables.* The components of Φ are *zero functions*.

Continuation is a computational method for successively growing a collection of solutions to a zero problem.

A zero problem is *adaptive* if Φ changes during continuation and *non-adaptive* otherwise.

The dimensional deficit of a zero problem is the difference n - m between the number of continuation variables and the number of zero functions. The dimensional deficit remains constant during continuation.

A solution u^* of a zero problem is *regular* if

 $\partial_u \Phi(u^*)$

has full rank. There exists a locally unique manifold of solutions to a zero problem through a regular solution. The manifold's dimension equals the corresponding dimensional deficit.

Solution manifolds of dimension 1 are called *branches*.

During continuation, the function $\Psi : \mathbb{R}^n \mapsto \mathbb{R}^r$ monitors properties of solutions to the zero problem. The components of Ψ are *monitor functions*.

The extended continuation problem

 $F(u,\mu)=0$

is defined in terms of the function

$$F:(u,\mu)\mapsto \left(egin{array}{c} \Phi(u)\ \Psi(u)-\mu\end{array}
ight)$$

and a vector of *continuation parameters* $\mu \in \mathbb{R}^r$. Its dimensional deficit equals n - m. If u^* is a regular solution of the zero problem, then $(u^*, \Psi(u^*))$ is a regular solution of the extended continuation problem.

Choose $\mathbb I$ and $\mathbb J$ such that $\mathbb I\cup\mathbb J=\{1,\ldots,r\}$ and $\mathbb I\cap\mathbb J=\emptyset.$ The restriction

$$G:(u,\mu_{\mathbb{J}})\mapsto F(u,\mu)\Big|_{\mu_{\mathbb{I}}=\mu_{\mathbb{I}}}$$

defines a restricted continuation problem

$$G(u,\mu_{\mathbb{J}})=0$$

with dimensional deficit $n - m - |\mathbb{I}|$.

If (u^*, μ_J^*) is a regular solution of the restricted continuation problem, then u^* is a regular solution of the *reduced continuation problem*

$$\left(egin{array}{c} \Phi(u) \ \Psi_{\mathbb{I}}(u) - \mu^*_{\mathbb{I}} \end{array}
ight) = 0$$

We construct a restricted continuation problem by defining the functions Φ and Ψ , and by choosing the index set \mathbb{I} corresponding to *inactive continuation parameters*.

We *initialize* a restricted continuation problem by assigning $\mu_{\mathbb{I}}^*$ and providing an initial solution guess for $(u, \mu_{\mathbb{J}})$ with the expectation that there exists a regular solution $(u^*, \mu_{\mathbb{J}}^*)$ nearby.

In practice, if u_0 is an initial solution guess for u, then

$$\mu_{\mathbb{I}}^* = \Psi_{\mathbb{I}}(u_0)$$

and $\Psi_{\mathbb{J}}(u_0)$ is an initial solution guess for $\mu_{\mathbb{J}}$.

In an adaptive continuation problem, the number and meaning of the zero functions and the continuation variables may change during continuation.

In contrast, the number and meaning of the monitor functions must remain unchanged also for adaptive continuation problems. They must be encoded accordingly.

We may explore different submanifolds of the solution manifold of the original zero problem by reassigning elements between \mathbb{I} and \mathbb{J} .

A reassignment from \mathbb{J} to \mathbb{I} imposes a *constraint* on the solutions to the zero problem. A reassignment from \mathbb{I} to \mathbb{J} *releases* the corresponding continuation parameter.

The Collocation Continuation Problem

For continuation of approximate solutions of the dynamical system

$$\dot{x} = F(t, x, p), (x, p) \in \mathbb{R}^n \times \mathbb{R}^q, t \in [T_0, T_0 + T],$$

define

$$\Phi: (\upsilon, T_0, T, p) \mapsto \left(\begin{array}{c} \frac{T}{2N} \operatorname{\mathfrak{vec}}(\kappa_F * F_{cn}) - W' \cdot \upsilon \\ Q \cdot \upsilon \end{array}\right)$$

in terms of the column matrix v of unknown values of the state variables on a mesh of N(m + 1) base points, and the values

$$F_{cn} = F(T_0 + Tt_{cn}, \mathfrak{vec}_n(W \cdot v), 1_{1,Nm} \otimes p)$$

of the vector field evaluated on a set t_{cn} of Nm collocation nodes on the interval [0, 1].

The Collocation Continuation Problem

The dimensional deficit of the collocation zero problem equals

n + q + 2.

For an autonomous vector field, append the monitor function

 $u \mapsto T_0$

and assign the index of the corresponding continuation parameter to $\mathbb{I}.$ The dimensional deficit of the corresponding restricted continuation problem then equals

$$n + q + 1$$
.

In COCO, we construct an empty continuation problem using the coco_prob command:

```
>> prob = coco_prob();
```

The commands

```
>> cat = @(x,p) [x(2,:); (1+x(2,:).^2)./x(1,:)];
>> t0 = 0;
>> x0 = [1 0];
>> prob = ode_isol2coll(prob, '', cat, t0, x0, []);
```

append the collocation zero problem on a default mesh consisting of 10 intervals with 5 base points and 4 collocation nodes in each interval, associate T_0 with the inactive continuation parameter 'coll.TO', and initialize the continuation problem with

$$v = \mathfrak{vec} \begin{pmatrix} \mathbf{1}_{1,50} \otimes \begin{pmatrix} \mathbf{1} & \mathbf{0} \end{pmatrix} \end{pmatrix}, \ T_0 = \mathbf{0}, \ T = \mathbf{0}, \ p = \emptyset.$$

The commands

```
>> data = coco_get_func_data(prob, 'coll', 'data');
>> maps = data.coll_seg.maps;
>> prob = coco_add_pars(prob, 'pars', ...
       [ maps.x0_idx; maps.x1_idx(1); maps.T_idx ], ...
       { 'y1s' 'y2s' 'y1e' 'T' });
```

append the monitor functions

$$\boldsymbol{\mu} \mapsto \left(\begin{array}{c} \boldsymbol{\upsilon}_i \\ \boldsymbol{\upsilon}_{f,1} \\ T \end{array} \right),$$

label the corresponding continuation parameters by 'y1s', 'y2s', 'y1e', and 'T', and assign the corresponding indexes to \mathbb{I} .

The dimensional deficit of the restricted continuation problem is now -1. The commands

```
>> cont_args = { 1, { 'T' 'y1e' }, [0 1] };
>> coco(prob, 'coll1', [], cont_args{:});
```

identify the desired manifold dimension as 1, reassign the indexes of the continuation parameters 'T' and 'y1e' to \mathbb{J} , and restrict continuation to the domain 'T' $\in [0,1]$.

By default, the collocation continuation problem is non-adaptive, so the number and meaning of the continuation variables and the zero functions is unchanged during continuation.

As an alternative, the commands

```
>> prob = coco_set(prob, 'cont', 'NAdapt', 10);
>> cont_args = { 1, { 'T' 'y1e' }, [0 1] };
>> coco(prob, 'coll1', [], cont_args{:});
```

instruct the continuation algorithm to make adaptive changes to the problem discretization after every ten successive steps of continuation.

Such adaptive changes are designed to ensure that a suitably estimated discretization error remains below a critical threshold during continuation.

More frequent changes, or a finer initial mesh, may be required in order to ensure successful continuation across the desired computational domain.