

RINO: some implementation notes

Sylvie Putot

No Institute Given

1 ODEs

We are given a system of ODEs

$$\dot{z}(t) = f(z, u, t)$$

with $z(0) = z^0$. Here, for efficiency reasons, we do not systematically rewrite u as a dimension of the system.

- `sysdim` is the dimension of the state space of the system of ODE
- `sysdim_params` is the dimension of the (possibly uncertain) parameters u that are not dimensions of the Jacobian
- `inputsdim` is the dimension of the uncertain parameters and inputs u that appear in the Jacobian
- when these parameters can be piecewise constant and not constant, `fullinputsdim` is the full dimension of these uncertain parameters and inputs
- `jacdim` = `sysdim` + `fullinputsdim`

`TM_Jac::build` and `TM_Jac::eval_Jac` build and evaluate the Taylor model for the Jacobian $J_{ij}(t, z_0) = \frac{\partial z_i}{\partial z_{0,j}}(t, z_0)$ and $J_{ij}(t, z_0) = \frac{\partial z_i}{\partial u_{0,j}}(t, z_0)$, as defined in Equation (17) of [HSCC2017]:

- in class `TM_Jac`, `J` corresponds to J_j in (17), and `J_rough` to R_{j+1} .
- in `TM_Jac::eval_Jac`, `Jaci` corresponds to $Jac_z f^{[i]}$ in (17)

The code is slightly more difficult to understand when `jacdim` is not equal to `sysdim`. Let us consider an example with dimension 2 and 2 parameters, the sensitivity matrix `J` we want to compute relies on products

$$Jac_z f^{[i]}([z_j]) \cdot [J_j]$$

of matrices of dimensions dimension (`jacdim` × `jacdim`) with, for example for $i = 1$,

$$Jac_z f^{[1]}([z_j]) \cdot [J_j] = \begin{pmatrix} \frac{\partial f_0}{\partial z_0} & \frac{\partial f_0}{\partial z_1} & \frac{\partial f_0}{\partial u_0} & \frac{\partial f_0}{\partial u_1} \\ \frac{\partial f_1}{\partial z_0} & \frac{\partial f_1}{\partial z_1} & \frac{\partial f_1}{\partial u_0} & \frac{\partial f_1}{\partial u_1} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} \frac{\partial z_0}{\partial z_0^0} & \frac{\partial z_0}{\partial z_1^0} & \frac{\partial z_0}{\partial u_0} & \frac{\partial z_0}{\partial u_1} \\ \frac{\partial z_1}{\partial z_0^0} & \frac{\partial z_1}{\partial z_1^0} & \frac{\partial z_1}{\partial u_0} & \frac{\partial z_1}{\partial u_1} \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

Thus, only part of the matrices are relevant. For efficiency, we store only the submatrices of dimension (`sysdim` × `jacdim`), and adapt the matrix operations such as matrix multiplication (`multJacfzJaczz0`) in that respect. We thus write

$$Jac_z f \cdot [J_j] = (Jac_z f \cdot Jac_{z^0} z \cdot Jac_z f \cdot Jac_u z + Jac_u^f)$$

This actually corresponds to Equations (8) and (9) (where β is z_0) of [HSCC2019].

2 Examples

Piecewise constant inputs Examples 19 and 21 are the following system:

$$\dot{x}(t) = f(u, t) = 2 + 2u + (1 - 2u)t$$

with $u \in [0, 1]$ and $x(0) = 0$. For now, time is embedded as a component of the system.

This example is a good example to test piecewise constant time-varying inputs. Let us take time horizon of 2 and stepsize of 1. If u is constant equal to u_0 on the whole interval, then

$$x(t) = \int_0^t 2 + 2u_0 + (1 - 2u_0)s ds = [(2 + 2u_0)s + (1 - 2u_0)s^2/2]_0^t$$

and in particular $x(2) = 6$. If u is only piecewise constant on each time setp of size 1, then

$$x(2) = \int_0^1 2 + 2u_1 + (1 - 2u_1)s ds + \int_1^2 2 + 2u_2 + (1 - 2u_2)s ds$$

and if both u_1 and u_2 are in $[0, 1]$ then $x(2) \in [5, 7]$.

On thix example, the maximal outer and inner approximations are exact. I thus added a non linearity to get non exact approximations in Example 22 which encodes

$$\dot{x}(t) = f(u, t) = 2 + 2(u^2 + u) + (1 - 2(u^2 + u))t$$

with $u \in [0, 1]$ and $x(0) = 0$.