

# Application of Input to State Stability to ecological reservoir models



Markus Müller, Carlos A. Sierra

<csierra@bgc-jena.mpg.de>

### Challenge

Many models in ecology and biogeochemistry, in particular models of the global carbon cycle, can be generalized as systems of nonautonomous ordinary differential equations (ODEs). For many applications, it is important to determine the stability properties for this type of systems, but most methods available for autonomous systems are not necessarily applicable for the non-autonomous case. We discuss here stability notions for non-autonomous nonlinear models represented by systems of ODEs explicitly dependent on time and a time-varying input signal. Is there a stability concept that is:

- 1. broad enough to encompass these models
- 2. rigorous enough to be proved analytically
- 3. interpretable in ecologically meaningful terms?

### Example: general soil model

 $\dot{\mathbf{C}} = \mathbf{I}(t) + \mathbf{T}(\mathbf{C}, t) \cdot \mathbf{N}(t, \mathbf{C}) \cdot \mathbf{C}(t)$ 

 $N_{i,i}(\mathbf{C},t) \geq 0 \quad \forall i$   $T_{i,i}(\mathbf{C},t) = -1 \quad \forall i$   $T_{i,j}(\mathbf{C},t) \geq 0 \quad \forall i \neq j$  $\sum_{i} T_{i,j}(\mathbf{C},t) = 1 \quad \forall j$ 

This model structure generalizes most SOM decomposition models with any arbitrary number of pools, including those describing nonlinear interactions among state variables. It enforces mass balance and substrate

dependence of decomposition,

4. Organic matter interactions

Examples for nonlinear models are:

- 1. Exoenzyme models [??]
- 2. AWB [?]
- 3. Bacwave [?]
- 4. MEND [?]
- 5. Manzoni [?]

2. ICBM [?]

3. RothC [??]

4. Century [?]

5. Formanic System Theory?

Also linear models fit into the

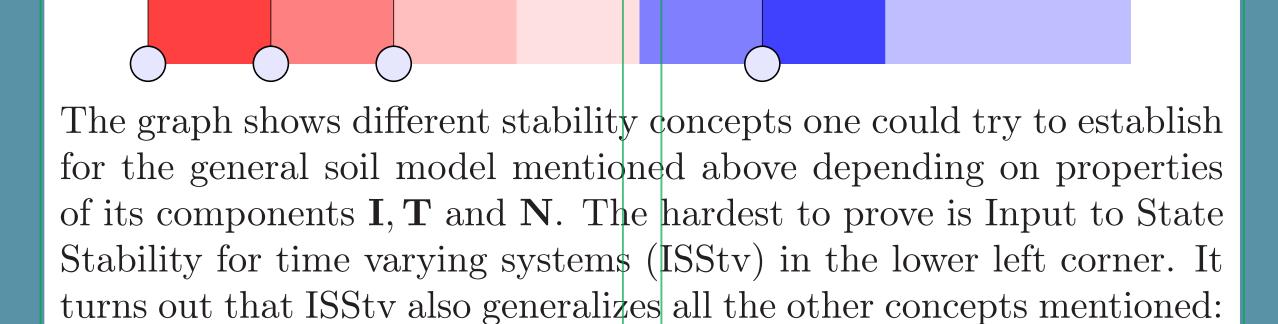
general framework

# Results I, ISS as generalization of available stability

- 1. Heterogeneity of decomposition rates
- 2. Transformations of organic
- matter
- variability 3. Environmental effects

is clearly desirable.

One general concept to encompass especially these nonlinear models



- In the case of Linear Time Invariant (LTI) systems ISS follows from the properties of the matrix. (eigenvalues)
- In the case of Linear Time Variant (LTV) systems it can be established if sufficient information about the state transition operator allows to prove uniform asymptotic stability UAS.
- For input free system (on the blue right-hand side) ISS reduces to Global Asymptotic Stability (GAS)

## Results II, ISS like behavior and proof for example system

The graphs show the reactions of a prototypical class of nonlinear two pool soil models to a disturbing time varying signal. This model is a technically simple place holder for ecologically motivated nonlinear systems like the soil models mentioned above to be analyzed in the future. It is given by:

fast enough". This situation seems to be typical: The problem of establishing ISS for e.g. all the I, T, N models based on the ecologic principles they follow, is too hard. But bio-chemical of biophysical restrictions could provide clues to ISS proofs for a particular system.

 $t_2 \qquad -||\mathbf{u}||_{\overline{\infty}}||\mathbf{u}||_{\infty} \qquad \qquad 0 \qquad \qquad ||\mathbf{u}||_{\infty}||\mathbf{u}||_{\infty}$ 

$$\dot{C}_x = I_x(t) - \left(C_x^2 + C_x\right) k_x(t) \tag{1}$$

$$\dot{C}_y = I_y(t) - \left(C_y^2 + C_y\right) k_y(t) \tag{2}$$

where  $C_x, C_y$  are the carbon contents of two unconnected pools and the bounded periodic functions  $k_x(t)$  and  $k_y(t)$  with:

$$k_{x_{min}} \le k_x(t) \le k_{x_{max}} \tag{3}$$

$$k_{y_{min}} \le k_y(t) \le k_{y_{max}} \tag{4}$$

describe the seasonal changes in decomposition speed. e.g.:

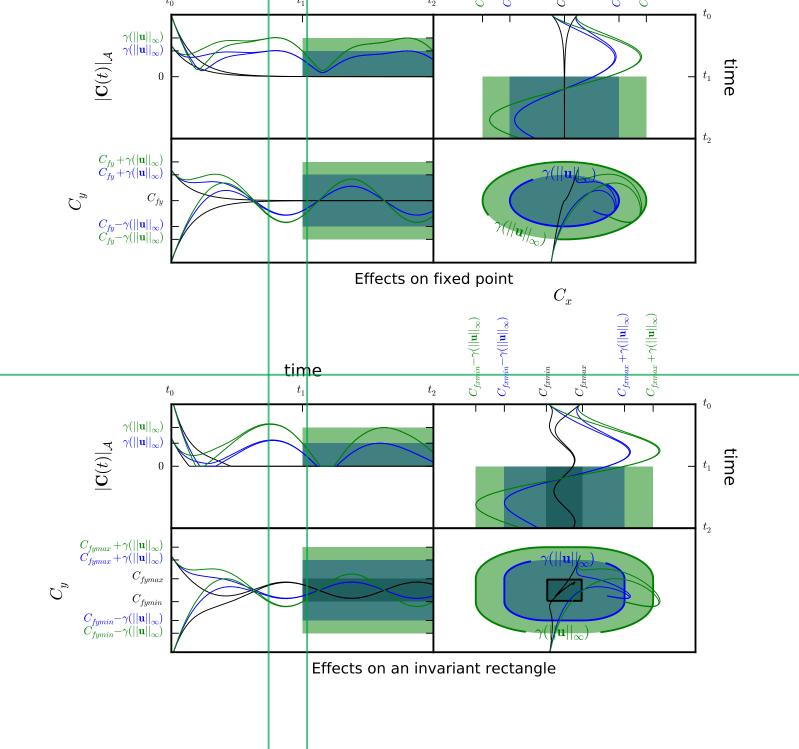
$$k_{x} = \frac{k_{xmax}}{2} + \frac{k_{xmin}}{2} + \frac{1}{2} (k_{xmax} - k_{xmin}) \sin(4t)$$
(5)  
$$k_{y} = \frac{k_{ymax}}{2} + \frac{k_{ymin}}{2} + \frac{1}{2} (k_{ymax} - k_{ymin}) \sin(4t)$$
(6)

The system can have a fixed point:

$$\mathbf{C}_f = \begin{pmatrix} C_{fx} \\ C_{fy} \end{pmatrix}$$

if the input streams have the same period and phase as the decomposition rates. For constant input streams it stays in a predictable region (an invariant set in the phase plane)

The plots show the typical behavior of an ISS system: The changes in the state variables will asymptotically converge to a region of stability around an invariant set, whose size is a monotone function of the size of the disturbance (denoted by  $|u|_{\infty}$ ). For this particular system we proved the ISS property rigorously. The proof relies on the construction of an ISS Lyapunov function whose choice is not determined but inspired by a property of the system interpretable in ecologically terms. Expressed casually: "The system can counteract supply changes



- 1. The four plots on the top show the disturbances.
- 2. The next four plots in the middle show the effect of this disturbances on the solutions for a system with fixed point.
- 3. The next four plots in the middle show the effect of this disturbances on the solutions for the 0.6 system which no longer has a fixed point, but at least an invariant set, the dark blue square in the middle.

#### Conclusion

We propose Input to State Stability (ISS) as candidate for the necessary generalization of the established analysis with respect to equilibria or invariant sets for autonomous systems, and showed for example systems its usefulness by applying it to reservoir models typical for element cycling in ecosystem, e.g. in soil organic matter decomposition. In a forthcoming paper we also showed how ISS generalizes existent concepts formerly only available for Linear Time Invariant (LTI) and Linear Time Variant (LTV) systems to the nonlinear case.

