

## 1 Example Applications of Pool Models

## 2 Reducing Model Complexity

- The carbon cycle as (de)motivating example
- Asking Simpler Questions
- Answer Questions more simply

## 3 Our proposed solution

- Definitions
- Algorithms
- Comparison to standard theory

## 4 Example

- Results

# Pool Models

A classification by required and obtainable information

Markus Müller, Holger Metzler, Carlos Sierra

August 5, 2019

- What are pool models?
- Why do we need them?
- What can they be used for?
  - What is needed?
  - What can we learn from them?

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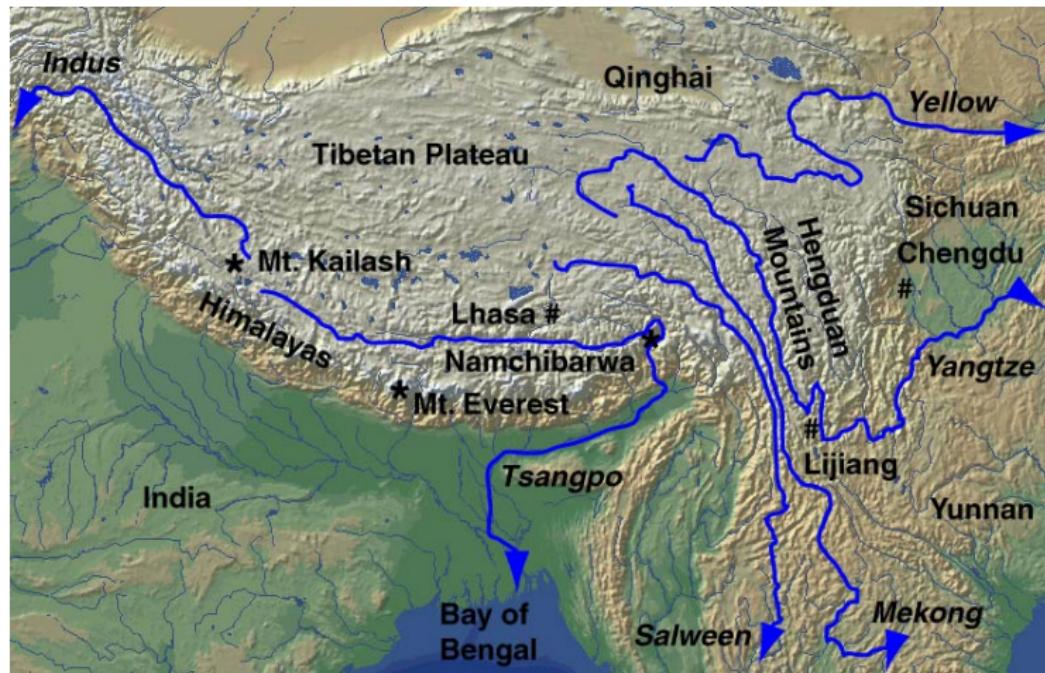
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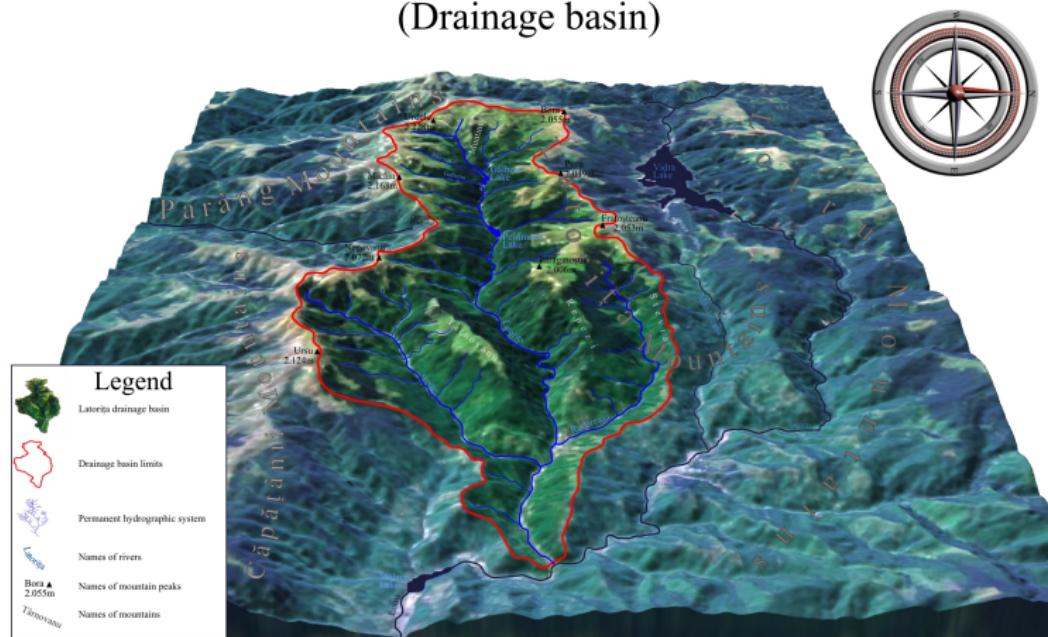
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# Hydrology Watersheds



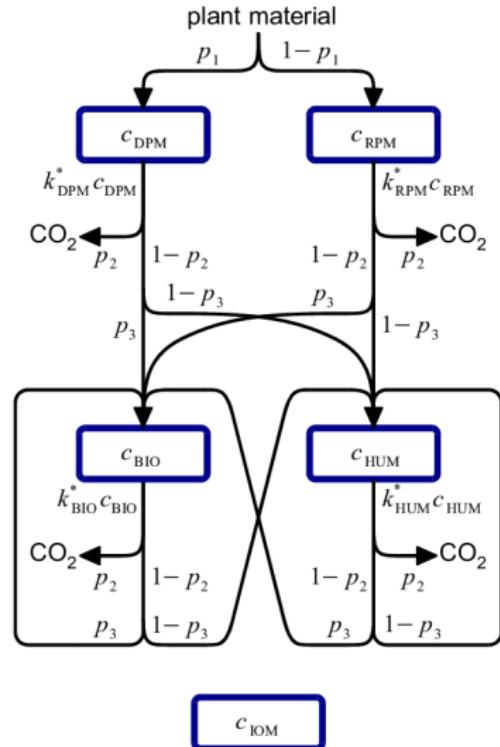
# Hydrology Catchments

## Latorița River, tributary of the Lotru River (Drainage basin)



# Plant Physiology/ Carbon allocation

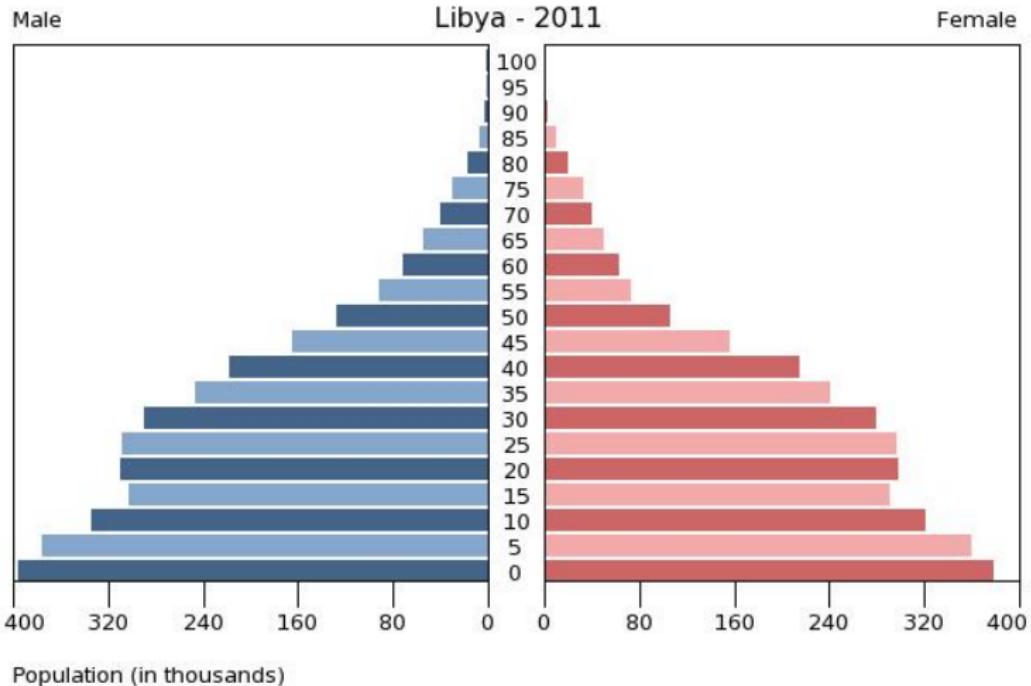
# Organic Matter Decomposition / Soil Models



# Ecosystem Models

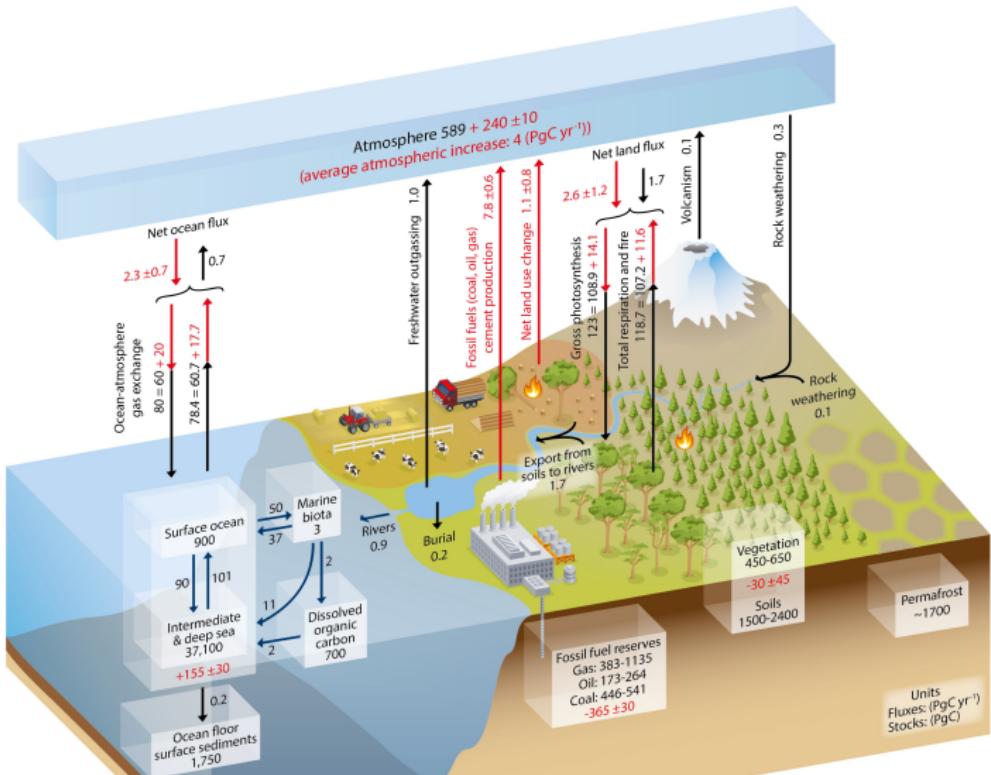
# Chemical Reactors

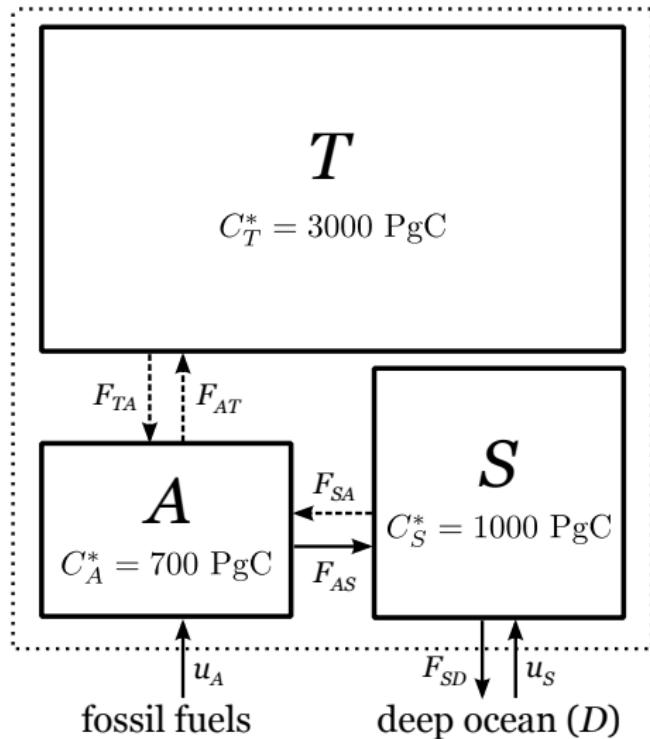
# Population Dynamics

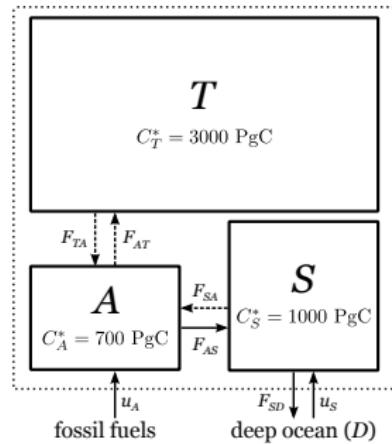
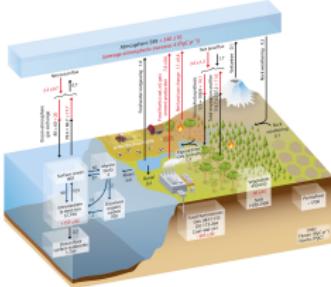






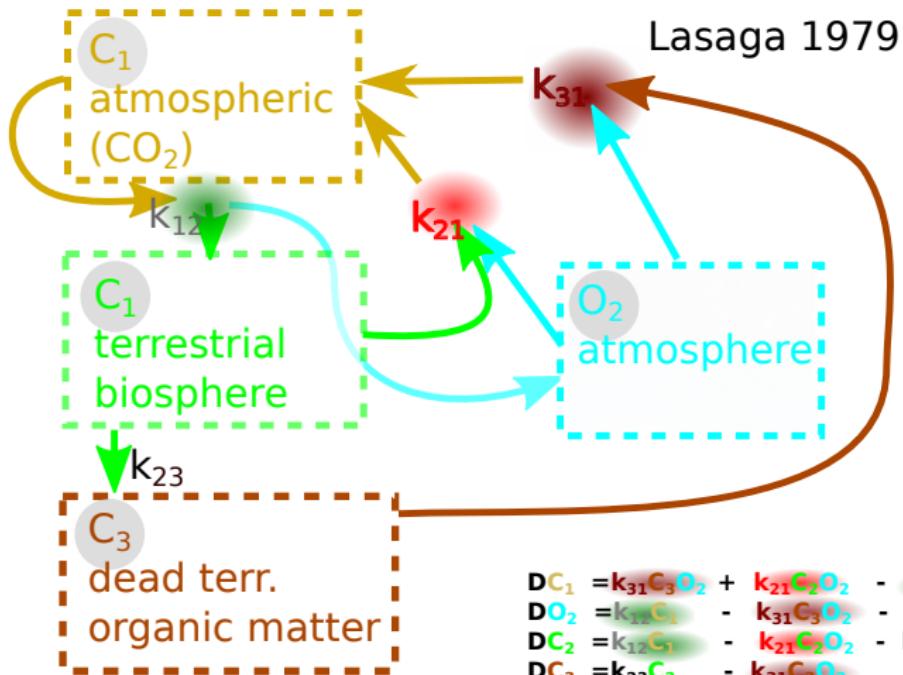




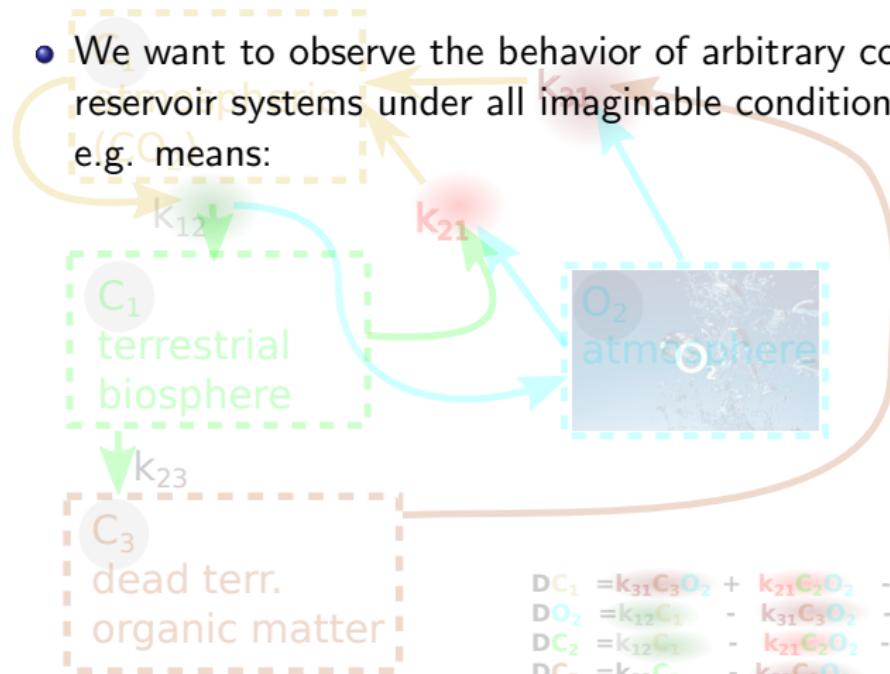




## Example Model

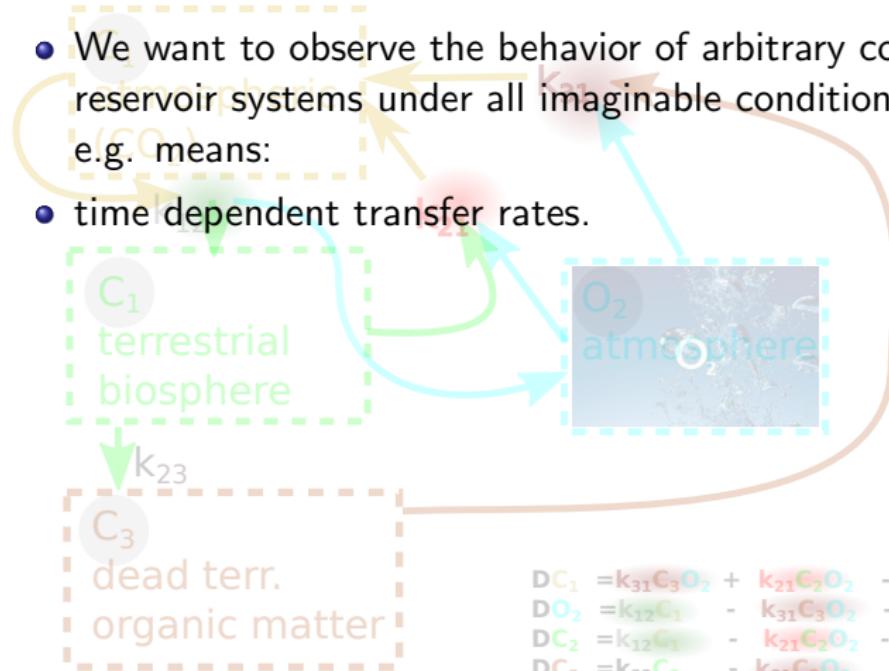


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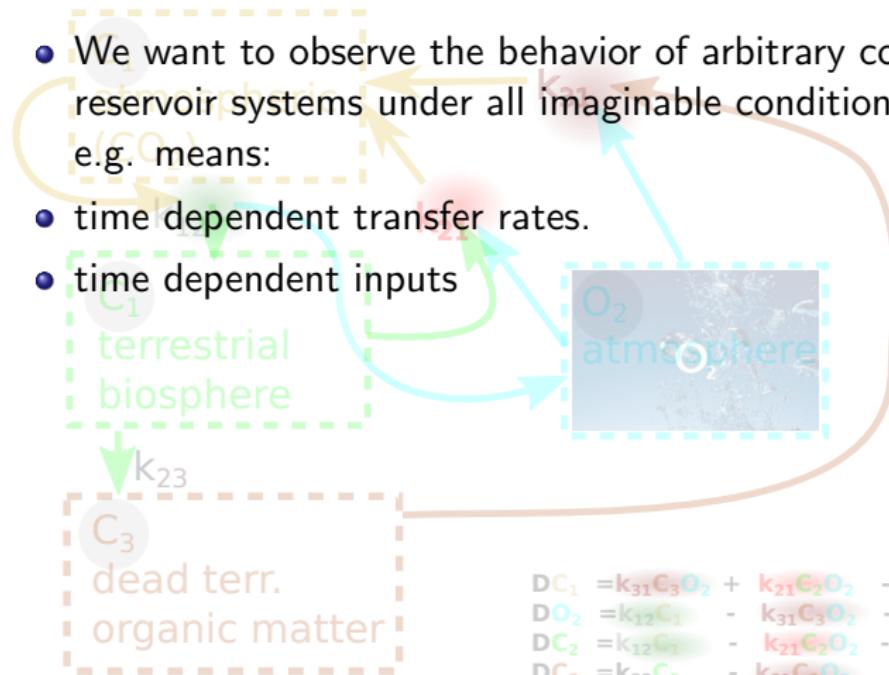
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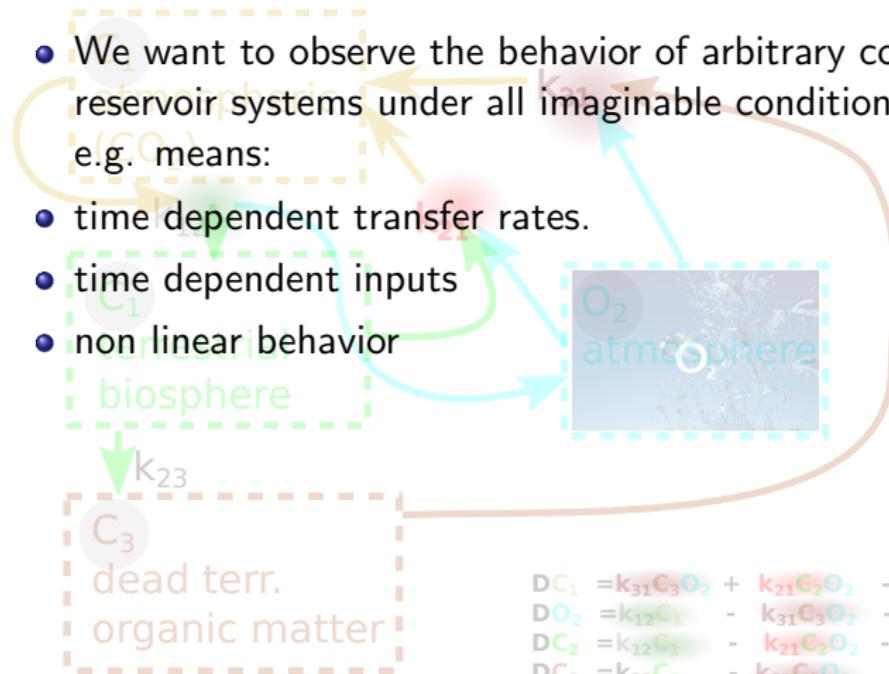
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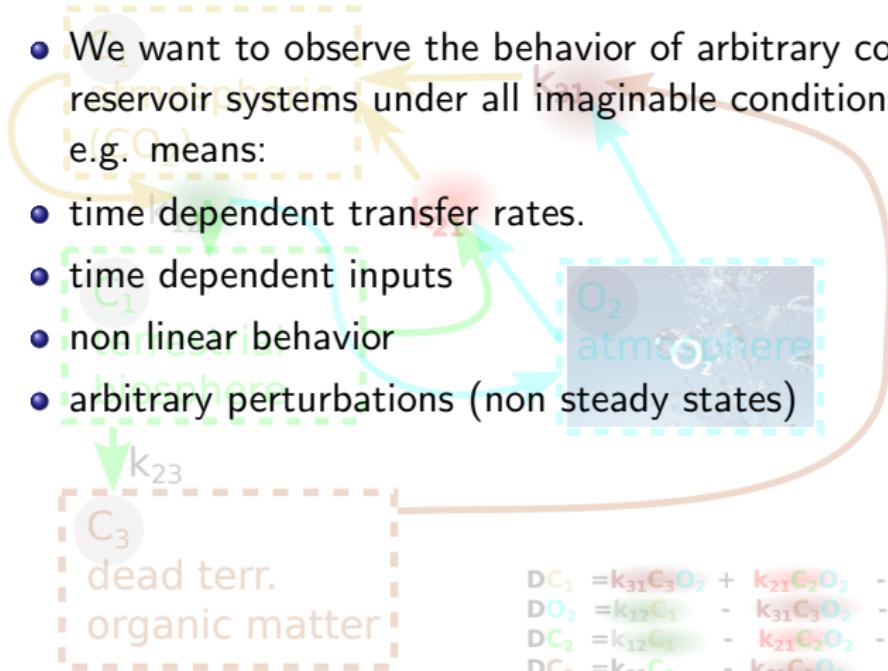
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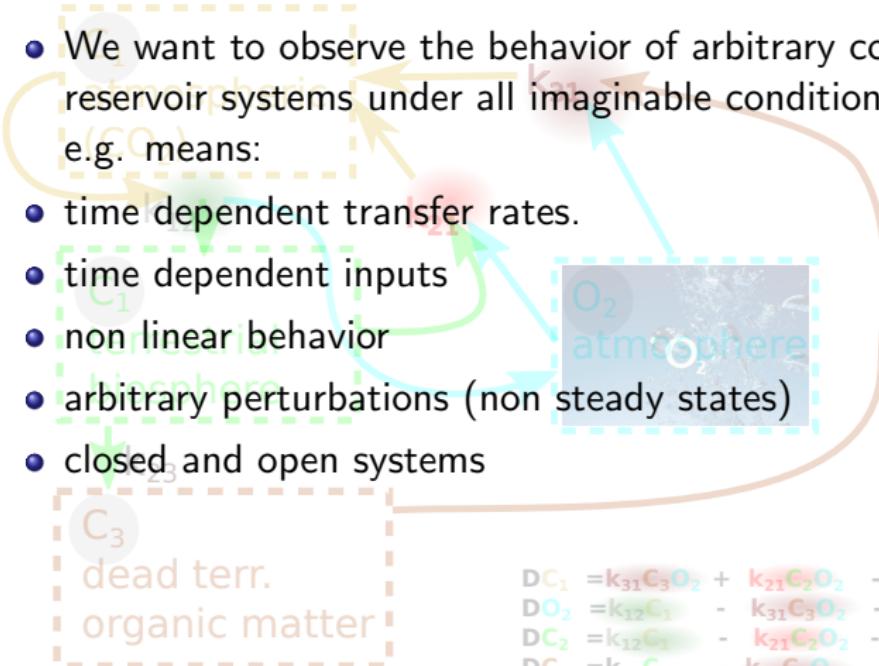
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  - reservoirs with nested reservoirs

dead terr.  
organic matter



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  - motivation for a numerical counterpart where all the assumptions (if any) should be **intuitive and obvious**.

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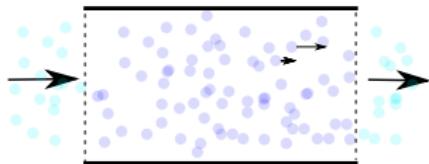
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→ no  $\infty$

# The general concept of a reservoir

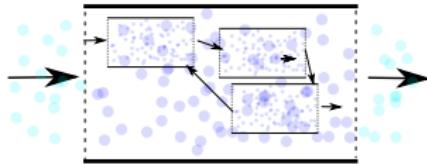


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# The general concept of a reservoir

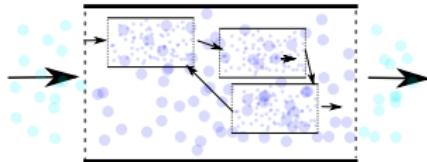
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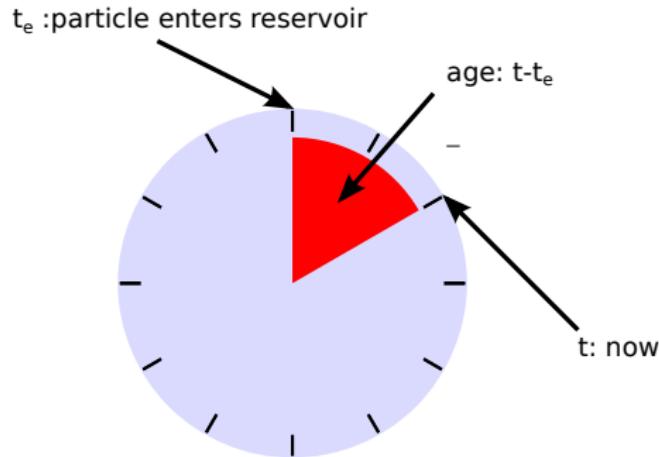
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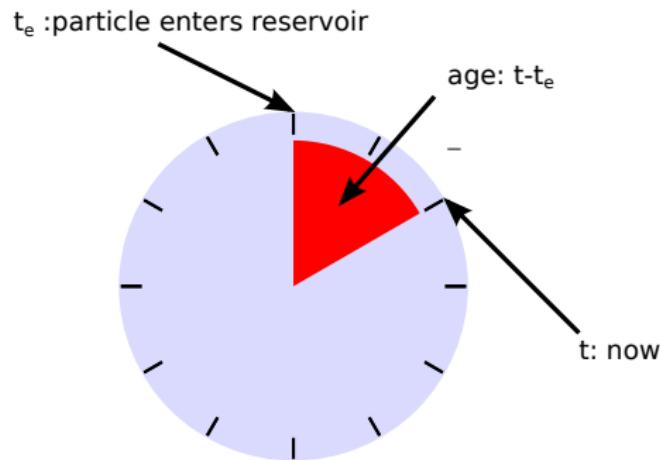


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- ④ reservoirs can be mixed or not

# age of a particle

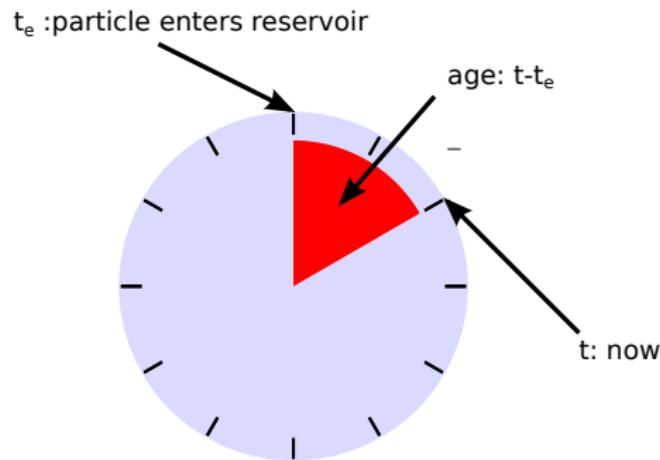


# age of a particle



- The “age ” is always defined in *context* of the reservoir

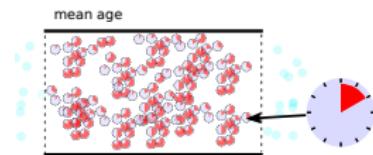
# age of a particle



- The “age” can not be negative!

# Mean age

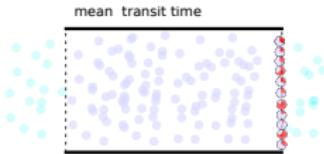
- Which set of particles to use for the average?  
proposition: *all* particles that are in the reservoir at the given time.  
→ usually depends on input rates as well as the dynamics of the system.



$$\bar{a}(t) = \frac{a_1 + a_2 + \cdots + a_N}{N}$$

With  $N = N(t)$  the number of all particles in the reservoir at time  $t$ .

# Mean transit time



$$\bar{t}_r(t) = \frac{a_1 + a_2 + \cdots + a_{n_o}}{n_o}$$

With  $n_o = n_o(t)$  the number of particles **just leaving** at time  $t$

- Can be time dependent as well

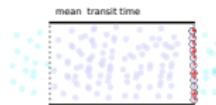
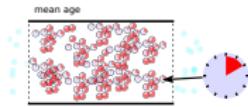
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- Can be time dependent as well
- Includes only the subset of particles that are just leaving at the given time. (Can only be computed when there is an output stream)

# Differences between mean age and mean transit time



- Includes **all** particles that are in the reservoir at the given time.
- Directly coupled to input rates
- Includes only the subset of particles that are **just leaving** at the given time.
- Indirectly coupled to inputs

# Iteration over all particles

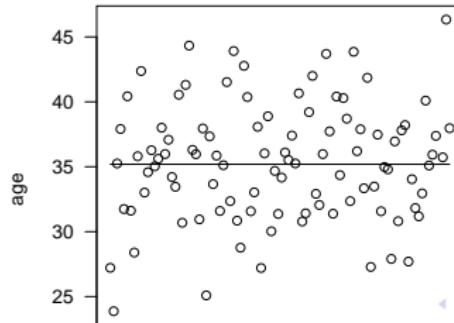
- ① To compute the mean transit time we have to identify the particles **just leaving**.
- ② Ask every leaving particle when it entered and compute its age.
- ③ Iterate over all particle and compute the average of their ages.

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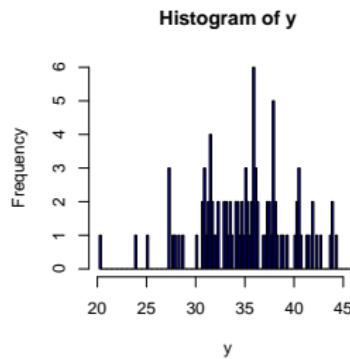
- ① as above
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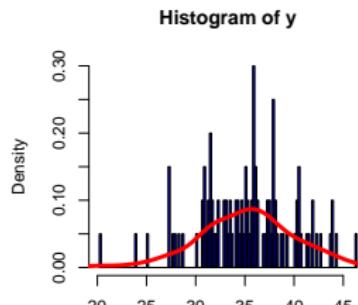
$$\bar{t}_r(t) = \frac{a_1 n_{a_1} + a_2 n_{a_2} + \cdots + a_n n_{a_n}}{n_o}$$

With  $n_o = n_o(t) = n_{a_1} + n_{a_2} + \cdots + n_{a_n}$



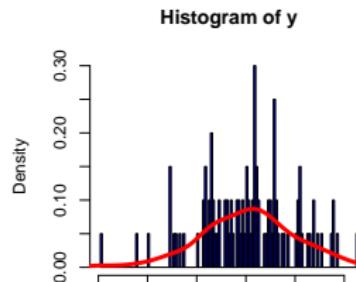
# Integration over a density

$$\begin{aligned}\bar{t}_r(t) &= \lim_{n \rightarrow \infty} \frac{a_1 n_{a_1} + a_2 n_{a_2} + \cdots + a_n n_{a_n}}{n_o} \\ &= \lim_{n \rightarrow \infty} \sum_{\text{minage}}^{\text{maxage}} a \frac{n(a)}{n_o} da \\ &= \int_{\text{minage}}^{\text{maxage}} a \psi(a) da\end{aligned}$$



## Same procedure for age density

$$\begin{aligned}\bar{a}(t) &= \lim_{n \rightarrow \infty} \frac{a_1 n_{a_1} + a_2 n_{a_2} + \cdots + a_n n_{a_n}}{n_p} \\ &= \lim_{n \rightarrow \infty} \sum_{\text{minage}}^{\text{maxage}} a \frac{n(a)}{n_p} da \\ &= \int_{\text{minage}}^{\text{maxage}} a \phi(a) da\end{aligned}$$



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- ④ for every  $t$  and every  $a : 0 < a < t$  compute  $\psi(a, t)$
- ⑤ for every  $t$  compute the expected value of  $a$  by integrating over the product:

$$E(t) = \int_0^t \psi(a, t) a \, da$$

# Transforming the system I

inject the solution into the operator

$$\begin{aligned}\dot{\vec{F}} &= \dot{\vec{C}}(\vec{C}, t) \\ &= \begin{pmatrix} \dot{F}_1(C_1, \dots, C_n, t) \\ \vdots \\ \dot{F}_n(C_1, \dots, C_n, t) \end{pmatrix} \\ &= \begin{pmatrix} \dot{I}_1(C_1, \dots, C_n, t) \\ \vdots \\ \dot{I}_n(C_1, \dots, C_n, t) \end{pmatrix} + \begin{pmatrix} \dot{O}_1(C_1, \dots, C_n, t) \\ \vdots \\ \dot{O}_n(C_1, \dots, C_n, t) \end{pmatrix}\end{aligned}$$

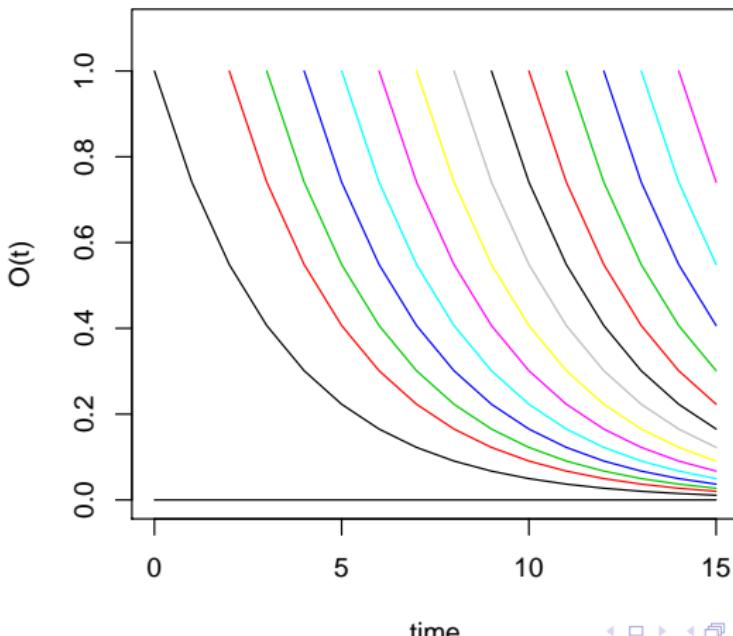
## Transforming the system II

For every component  $F_i$  do:

$$\begin{aligned}\dot{F}_i(t) &= \frac{I_i(C_1(t), \dots, C_n(t), t)}{C_i(t)} C_i(t) - \frac{O_i(C_1(t), \dots, C_n(t), t)}{C_i(t)} C_i(t) \\ &= I_{i_{lin}}(t) C_i(t) - O_{i_{lin}}(t) C_i(t) \\ &= F_{i_{lin}}(t) C_i(t)\end{aligned}$$

Now the equations are uncoupled and linear.

# Accumulating previous inputs



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- ⑤ compute the derivative with respect to  $a$

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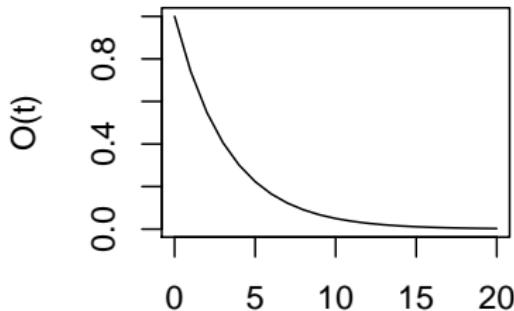
- Input(rate): impulsive only at the start  $i = C_0\delta(0)$
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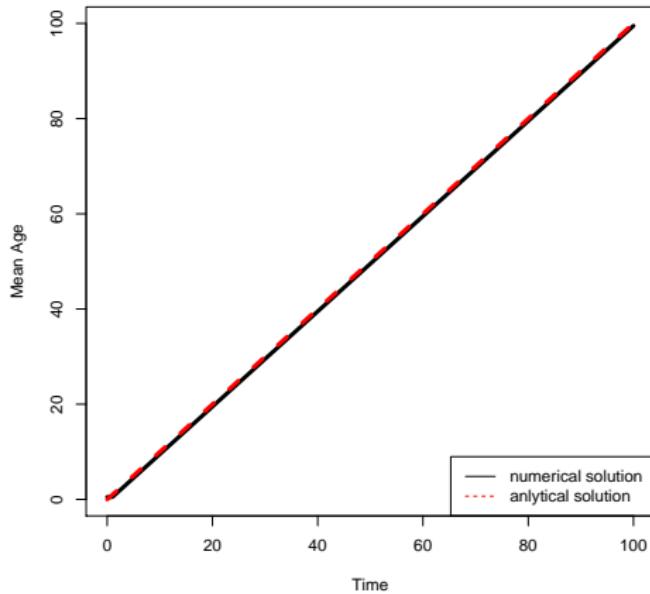
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- solution:

## Conceptual example

- Input(rate): impulsive only at the start  $i = C_0\delta(0)$
- Output(rate):  $\dot{O}(C, t) = -kC(t)$
- wanted:  $C(t), \phi(t), \psi(t), \bar{t}_r, \bar{a}$
- solution:
  - $C(t) = C_0 e^{-kt}$



# Intuitive solution for the example



## Theory for the example

Manzoni, Katul, Porporato 2009:

"for any linear systems the transit time distribution is the output flow resulting from an impulsive unitary input."

$$\begin{aligned} O(t) &= \int_0^{\infty} \psi(T) I(t - T) dT \\ &= \psi(t) \text{ for } I = \delta(t - T) \\ &= e^{-kt} \end{aligned}$$

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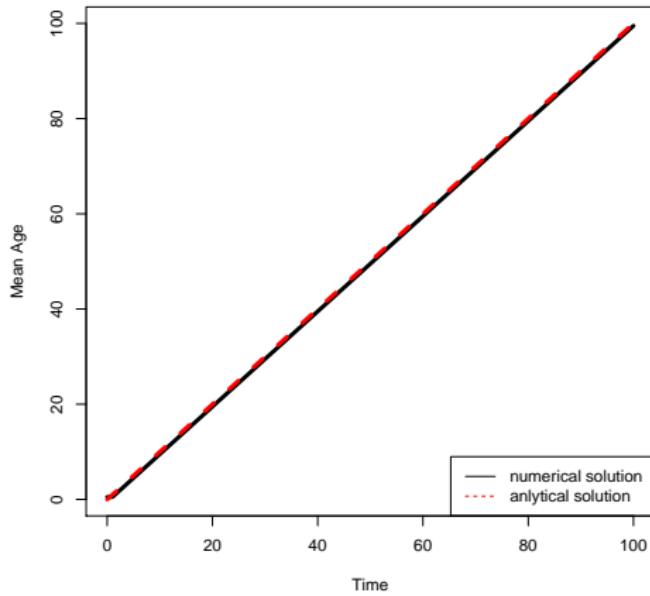
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# Intuitive Solution to the Example

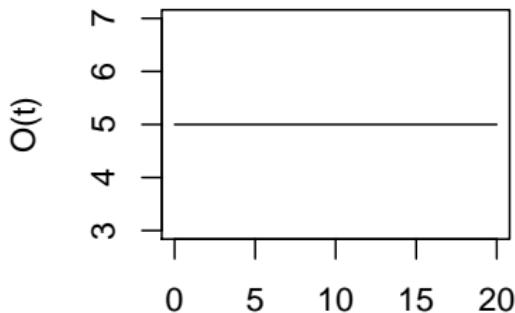


## Conceptual example II

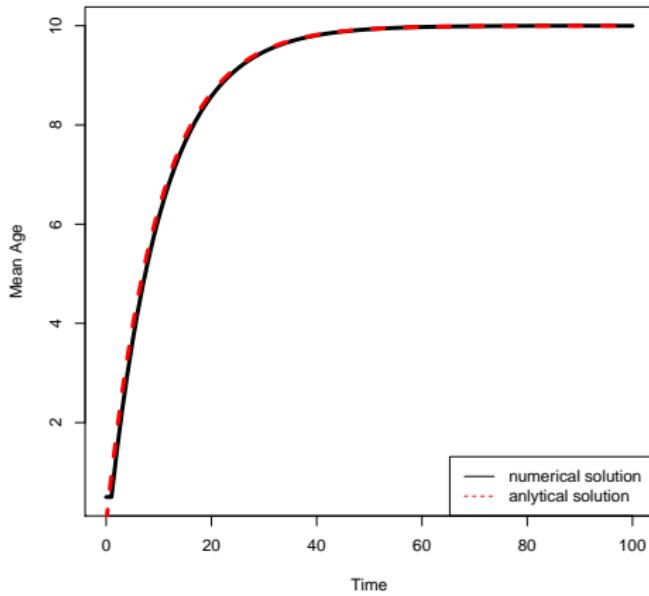
- Input(rate): constant  $I_0 = C_0 k$
- Output(rate):  $\dot{O}(C, t) = -kC(t)$
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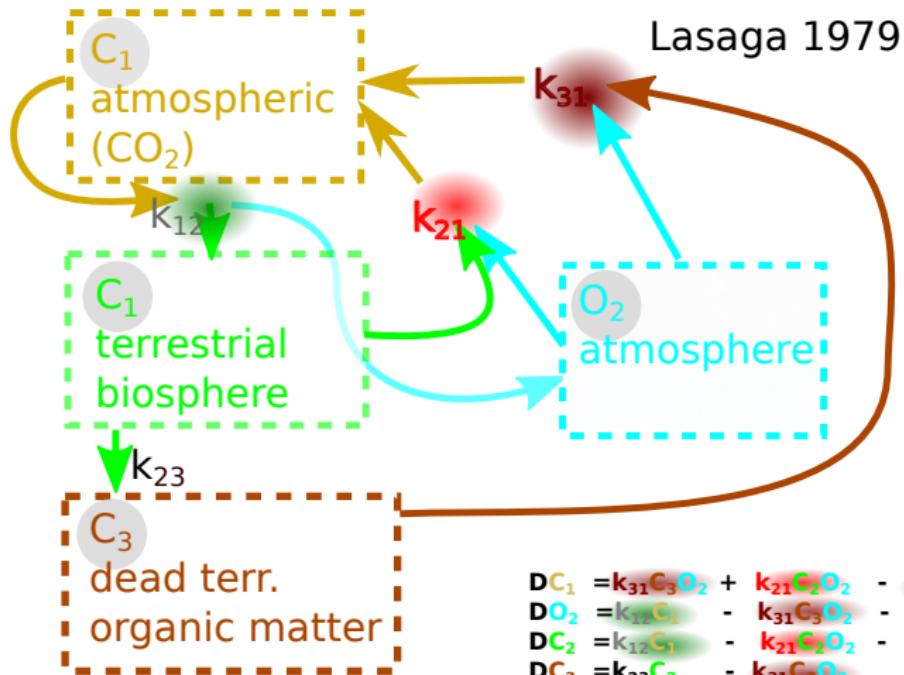
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- solution:
  - $C(t) = C_0$



# Intuitive Solution to the Example in steady state

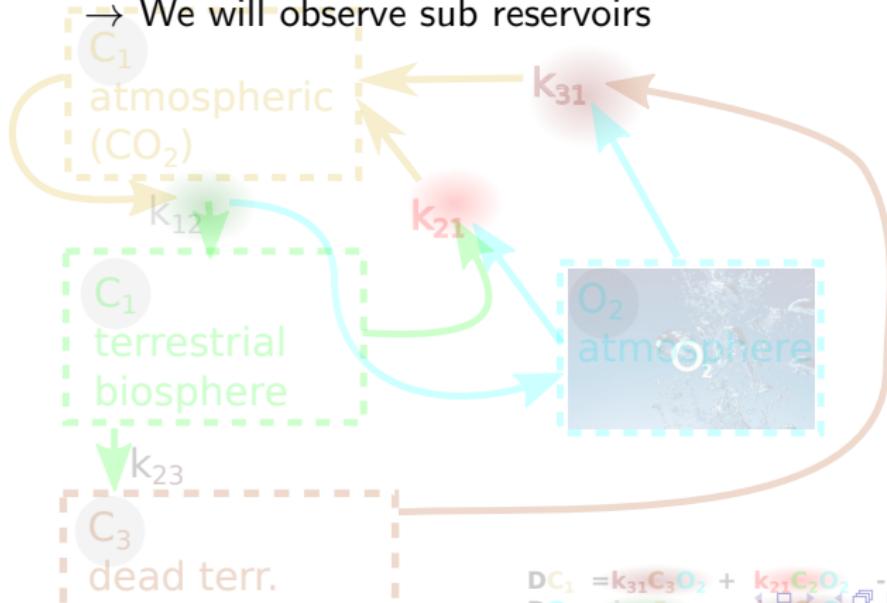


## Example Model



# Didactical Considerations

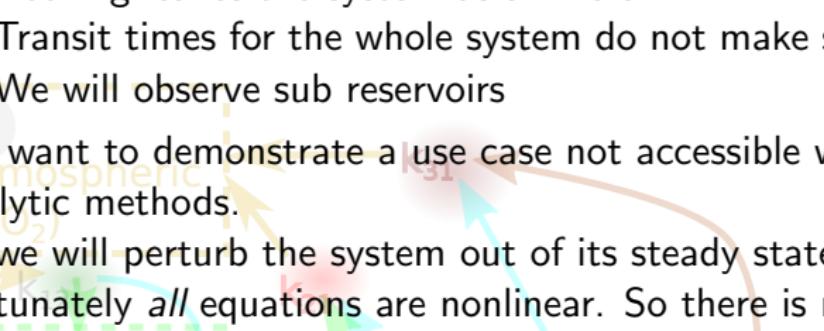
- The system is a *cycle*.
  - Nothing leaves the system as a whole.
  - Transit times for the whole system do not make sense.
  - We will observe sub reservoirs

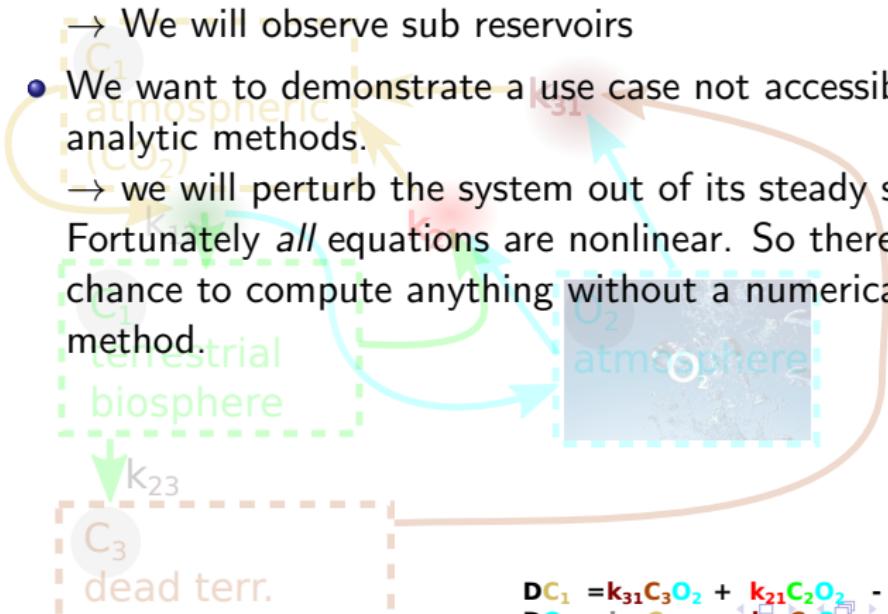


## Didactical Considerations

- The system is a *cycle*.
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    - We will observe sub reservoirs
  - We want to demonstrate a use case not accessible with analytic methods.
    - we will perturb the system out of its steady state.

Fortunately *all* equations are nonlinear. So there is no chance to compute anything without a numerical method.



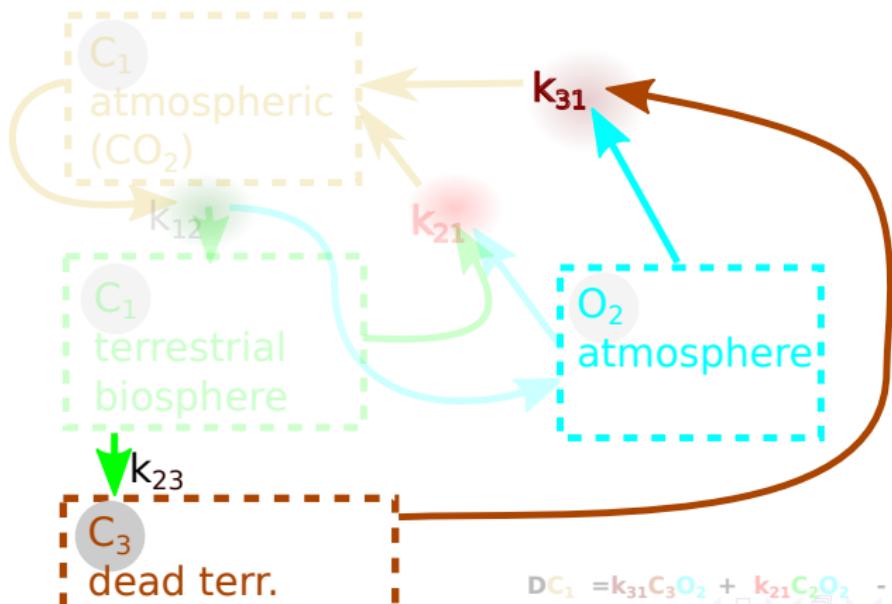


$$\begin{aligned} \text{DC}_1 &= k_{31}C_3O_2 + k_{21}C_2O_2 - k_{12}C_1 \\ \text{DO}_2 &= k_{12}C_1 - k_{31}C_3O_2 - k_{21}C_2O_2 \end{aligned}$$

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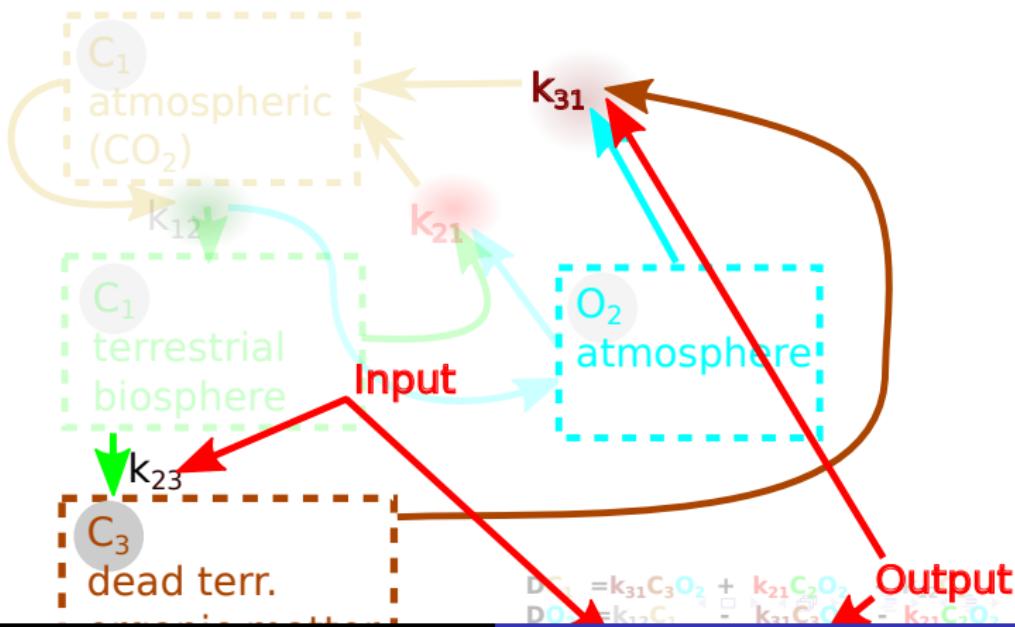
- The system is a *cycle*.
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  - We will observe sub reservoirs
- We want to demonstrate a use case not accessible with analytic methods.
  - we will perturb the system out of its steady state.
  - Fortunately *all* equations are nonlinear. So there is no chance to compute anything without a numerical method.
- We do not want to complicate matters unnecessarily.
  - We will go for the simplest equation.
  - observe the last pool.

# Didactical Considerations

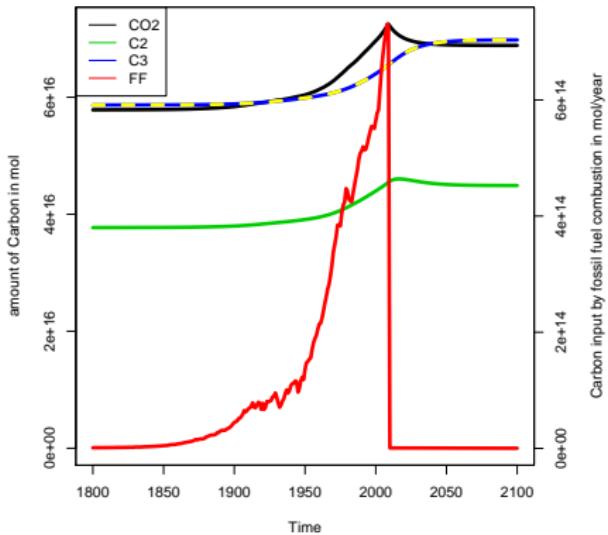


$$\begin{aligned} \Delta C_1 &= k_{31}C_3O_2 + k_{21}C_2O_2 - k_{12}C_1 \\ \Delta O_2 &= k_{12}C_1 - k_{21}C_2O_2 - k_{31}C_3O_2 \end{aligned}$$

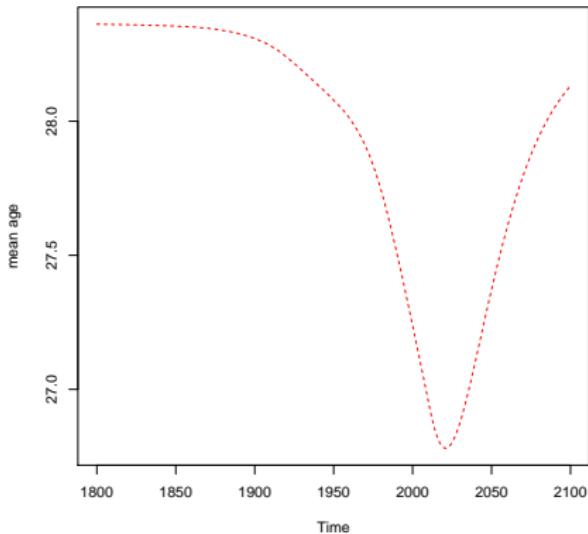
# Didactical Considerations



# Solution



# Mean age as function of time



# Conclusions and outlook

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- The previously available methods for the computation of mean ages and mean transit times may lead to confusing interpretations if applied to time invariant systems.
- One more generally applicable set of definitions has been proposed.
- A numerical method has been (partly) developed to obtain the desired values for nonlinear, coupled, time variant systems.
- The method has to be applied to coupled open systems
- Other definitions of mean age and transit times could be evaluated for suitability.

Thank you for your attention

The image shows a man with glasses and a bow tie standing in front of a chalkboard. The chalkboard is filled with mathematical equations, including integrals and logarithmic functions. The equations appear to be related to the topic of pool models and reducing model complexity.

$$I = \int \sqrt{1+(1+u^2)^2} \frac{u du}{1+u^2}, z=1+u^2, dz$$
$$I = \frac{1}{2} \int \sqrt{1+z^2}$$
$$\int \frac{1+z^2}{z \sqrt{1+z^2}} dz = \frac{1}{2} \sqrt{1+z^2} + \frac{1}{2} I_1, I_1 = \int \frac{-dt}{t \sqrt{1+\frac{1}{t^2}}} = -\ln(t + \sqrt{1+\frac{1}{t^2}}) + C_1$$
$$= \ln z - \ln(1 + \sqrt{1+z^2}) + C_1 = \ln(\frac{z}{1 + \sqrt{1+z^2}}) + C_1$$
$$= \ln(1 + \tan^2 x) - \ln(1 + \sqrt{1 + (\tan^2 x)}) + C_1$$
$$= \ln(\cos^2 x + \sqrt{\cos^4 x + 1}) + 2 \ln |\cos x| + C_1$$