

The stoich module

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Note: this is the stoich.ipynb notebook. The PDF version “The stoich module” is available [here](#).

1 Introduction

BondGraphTools is a python based toolkit for the creation and analysis of bond graph models of physical systems. Such physical systems include biomolecular systems; **stoich** is a toolkit for the stoichiometric analysis of such systems.

This document provides a simple introduction to **stoich** by means of a built-in bond graph model of a simple enzyme-catalysed reaction.

2 Example.

stoich.model() implements the enzyme-catalysed reaction: $A + E = C = B + E$ where A is the substrate, B the product, E the enzyme and C an intermediate compound. This can be analysed using the following code.

First import some code:

```
[1]: import BondGraphTools as bgt
import numpy as np
import sympy as sp
import IPython.display as disp
import stoich as st
```

2.1 Basic analysis

Now perform stoichiometric analysis on the model:

```
[2]: s = st.stoich(st.model())
```

```
Computing N ...
Swapping Re:r1 for two Sf in ABCE
Swapping Re:r2 for two Sf in ABCE
Done.
Computing K ...
Done.
Computing G ...
Done.
```

s is a Python dict containing the stoichiometric information. For example, the stoichiometric matrix N where $\dot{X} = NV$ is revealed as:

```
[3]: print(s['N'])
```

```
[[ -1  0]
 [ 0  1]
 [ 1 -1]
 [-1  1]]
```

This can be displayed in a more readable form as:

```
[4]: disp.Latex(st.sprintl(s, 'N'))
```

[4]:

$$N = \begin{pmatrix} -1 & 0 \\ 0 & 1 \\ 1 & -1 \\ -1 & 1 \end{pmatrix} \quad (1)$$

The state (vector of concentrations) X and the vector V of reaction flows are:

```
[5]: disp.Latex(st.sprintl(s, 'species'))
```

[5]:

$$X = \begin{pmatrix} X_A \\ X_B \\ X_C \\ X_E \end{pmatrix} \quad (2)$$

```
[6]: disp.Latex(st.sprintl(s, 'reaction'))
```

[6]:

$$V = \begin{pmatrix} V_{r1} \\ V_{r2} \end{pmatrix} \quad (3)$$

The corresponding reactions can be displayed:

```
[7]: disp.Latex(st.sprintl(s))
```

[7]:



The stoichiometric matrix N gives the species state X in terms of reaction flow V from $\dot{X} = NV$. N is also used together with thermodynamic constants K and rate constants κ to give an explicit expression for reaction flow V in terms of species state X . **stoich** computes the symbolic expression as:

```
[8]: disp.Latex(st.sprintl(s, 'N'))
```

[8]:

$$N = \begin{pmatrix} -1 & 0 \\ 0 & 1 \\ 1 & -1 \\ -1 & 1 \end{pmatrix} \quad (6)$$

```
[9]: disp.Latex(st.sprintv1(s))
```

[9]:

$$v_{r1} = \kappa_{r1} (K_A K_E x_A x_E - K_C x_C) \quad (7)$$

$$v_{r2} = \kappa_{r2} (-K_B K_E x_B x_E + K_C x_C) \quad (8)$$

2.2 Conserved moieties and Pathways

Conserved moieties are revealed by the matrix G where $G^T N = 0$. In this case:

```
[10]: disp.Latex(st.sprintl(s, 'G'))
```

```
[10]:
```

$$G = \begin{pmatrix} 1 & 1 & 1 & 0 \\ -1 & -1 & 0 & 1 \end{pmatrix} \quad (9)$$

The first row corresponds to $\dot{x}_A + \dot{x}_B + \dot{x}_C = 0$, the sum of the rows $(0 \ 0 \ 1 \ 1)$ corresponds to $\dot{x}_C + \dot{x}_E = 0$

Pathways are revealed by the matrix K where $NK = 0$. In this case:

```
[11]: disp.Latex(st.sprintl(s, 'K'))
```

```
[11]:
```

$$K \Rightarrow (\quad) \quad (10)$$

There are no pathways: there is zero flow ($V = 0$) in the steady state.

2.3 Chemostats.

Consider the case where both substrate A and product B are chemostats:

```
[12]: chemostats = ['A', 'B']
```

The same system, but with the chemostats, can be analysed using:

```
[13]: sc = st.statify(s, chemostats=chemostats)
```

The stoichiometric matrix N is now:

```
[14]: disp.Latex(st.sprintl(sc, 'N'))
```

```
[14]:
```

$$N = \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 1 & -1 \\ -1 & 1 \end{pmatrix} \quad (11)$$

The first two rows are zero, corresponding to $\dot{x}_A = \dot{x}_B = 0$: this is because both substrate A and product B are chemostats.

The pathway matrix K is now:

```
[15]: disp.Latex(st.sprintl(sc, 'K'))
```

```
[15]:
```

$$K = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad (12)$$

This means that the flow through reactions r1 and r2 are the same and can be non-zero at steady-state. The conserved moieties of this chemostated system are revealed by the matrix G

```
[16]: disp.Latex(st.sprintl(sc, 'G'))
```

```
[16]:
```

$$G = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{pmatrix} \quad (13)$$

The three rows correspond to: - $\dot{x}_A = 0$ (x_A is constant) - $\dot{x}_B = 0$ (x_B is constant) - $\dot{x}_C + \dot{x}_E = 0$ ($x_C + x_E$ is constant)

2.4 Pathway analysis

```
[17]: ## Find the pathway stoichiometric matrix
      sp = st.path(s,sc)
      ## And show the coreponding reaction
      disp.Latex(st.sprintl(sp))
```

```
[17]:
```



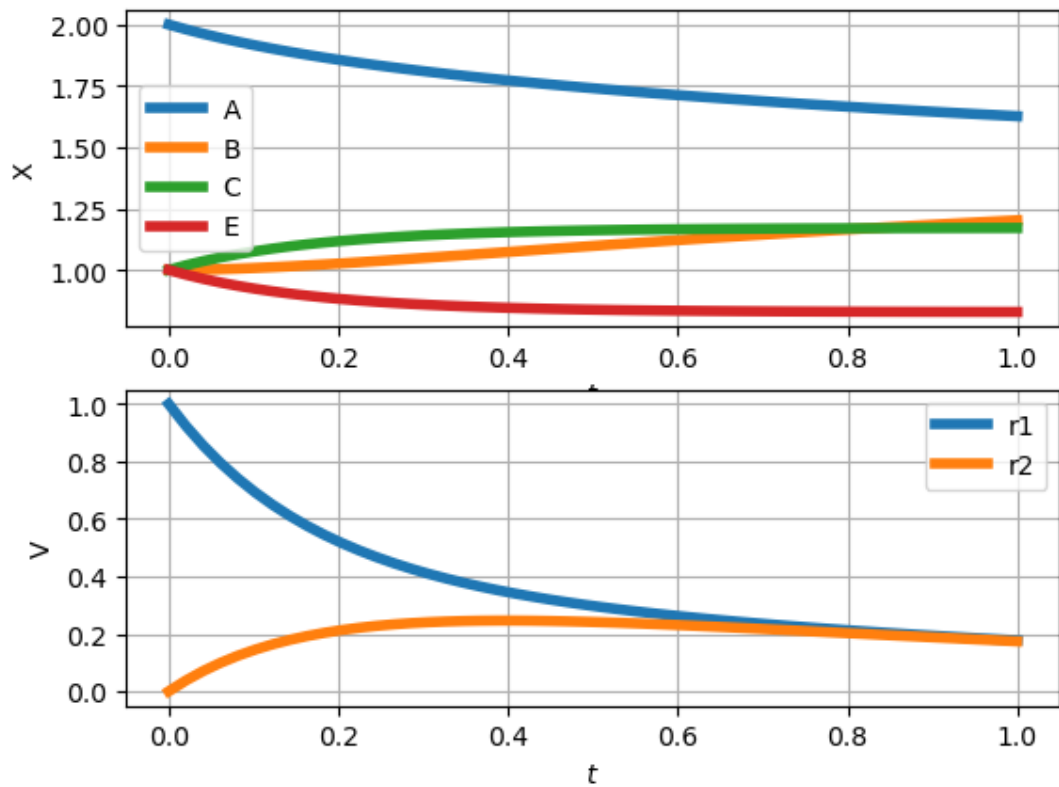
2.5 Simulation

Although **BondGraphTools** has its own simulation tool, the particular form of stoichiometric equations allows for a special purpose simulation tool taking advantage of explicit equations and reducing the state dimension in the presence of conserved moieties.

The system (without chemostats) can be simulated as:

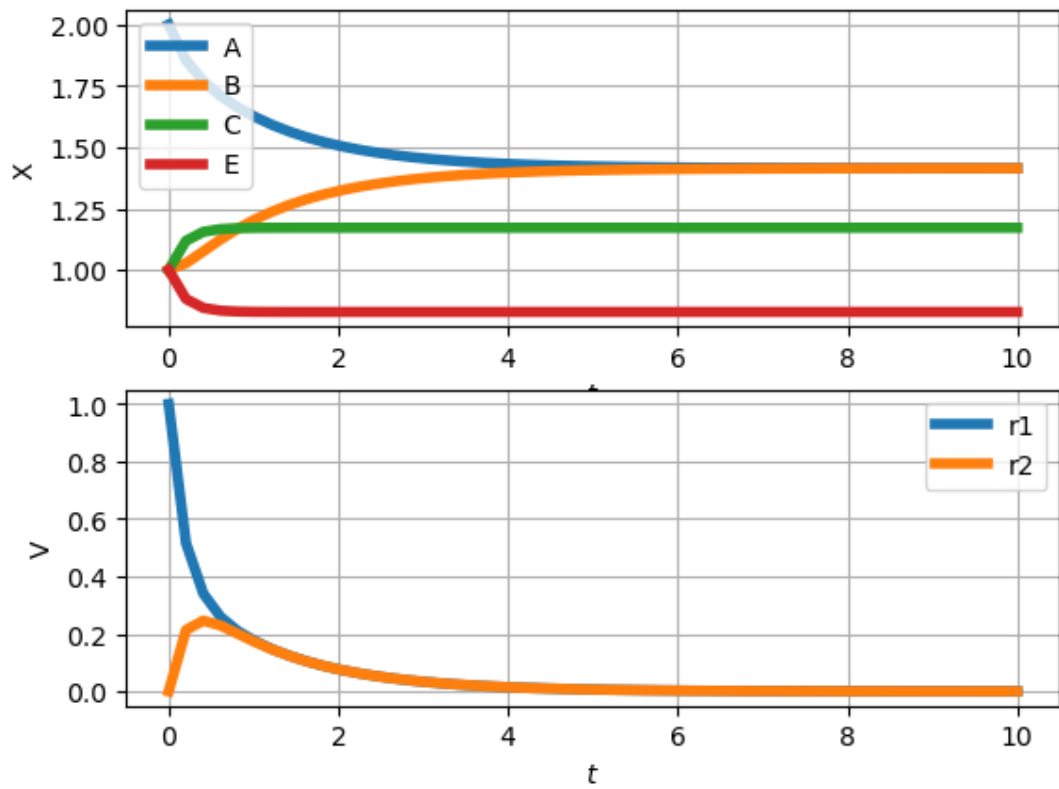
```
[18]: X0 = np.array([2,1,1,1]) # Set initial states
      result = st.sim(s,X0=X0) # Simulate
```

```
[19]: st.plot(s,result)
```



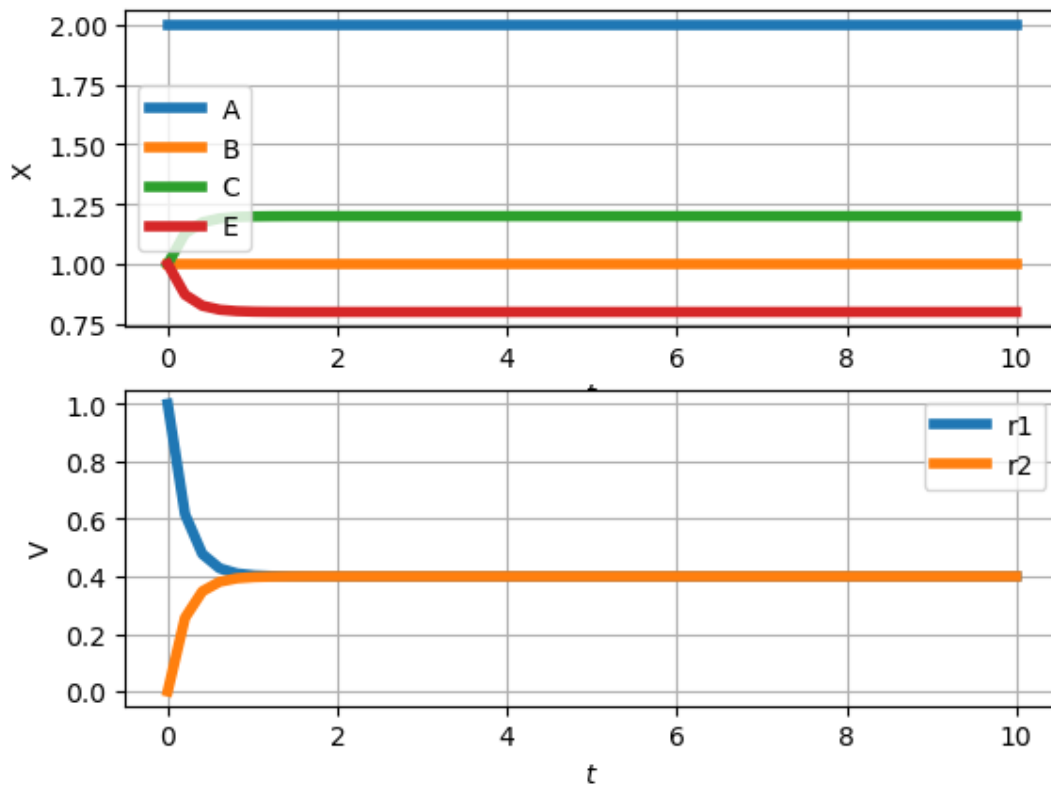
The flows V through $r1$ and $r2$ seem to be heading towards zero as predicted by pathway analysis. This can be verified by simulating over a longer time:

```
[20]: t = np.linspace(0,10)
      result = st.sim(s,X0=X0,t=t)
      st.plot(s,result)
```



The system with chemostats can be simulated as:

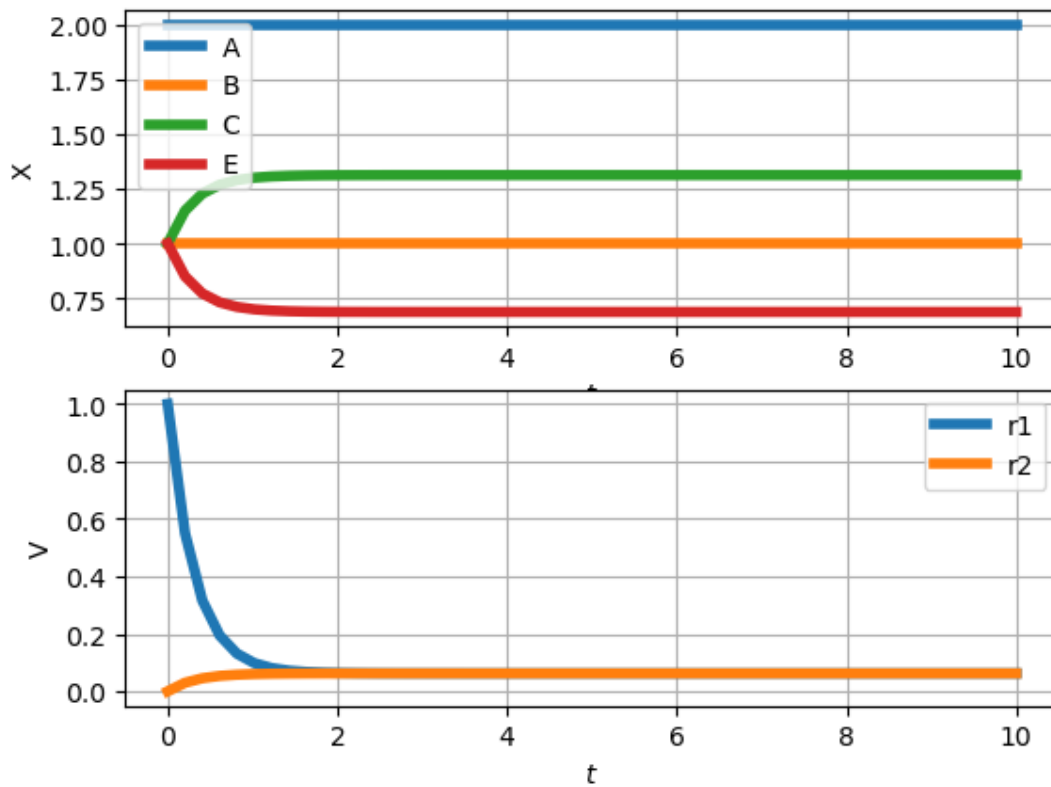
```
[21]: t = np.linspace(0,10)
      result = st.sim(s,X0=X0,t=t,sc=sc)
      st.plot(s,result)
```



As predicted by pathway analysis, the two flows converge on a non-zero value.

Of course, these simulations have been using default (unity) values for parameters. These defaults can be changed by explicitly supplying parameters:

```
[22]: parameter={'kappa_r2':0.1}
      result = st.sim(s,X0=X0,t=t,sc=sc,parameter=parameter)
      st.plot(s,result)
```

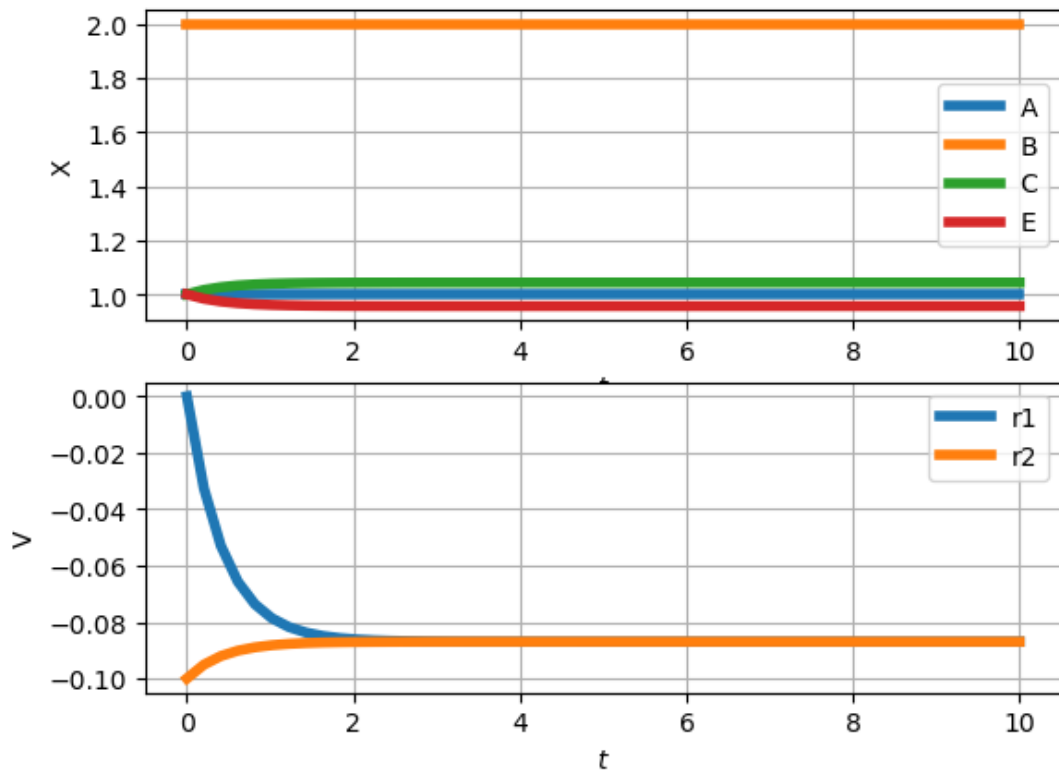
Moreover, the four initial conditions can be explicitly chosen, for example: -

$$X_0 = (1 \quad 2 \quad 1 \quad 1)^T$$

Note that the default value was -

$$X_0 = (2 \quad 1 \quad 1 \quad 1)^T$$

```
[23]: X0 = np.array([1,2,1,1])
      result = st.sim(s,t=t,sc=sc,parameter=parameter,X0=X0)
      st.plot(s,result)
```

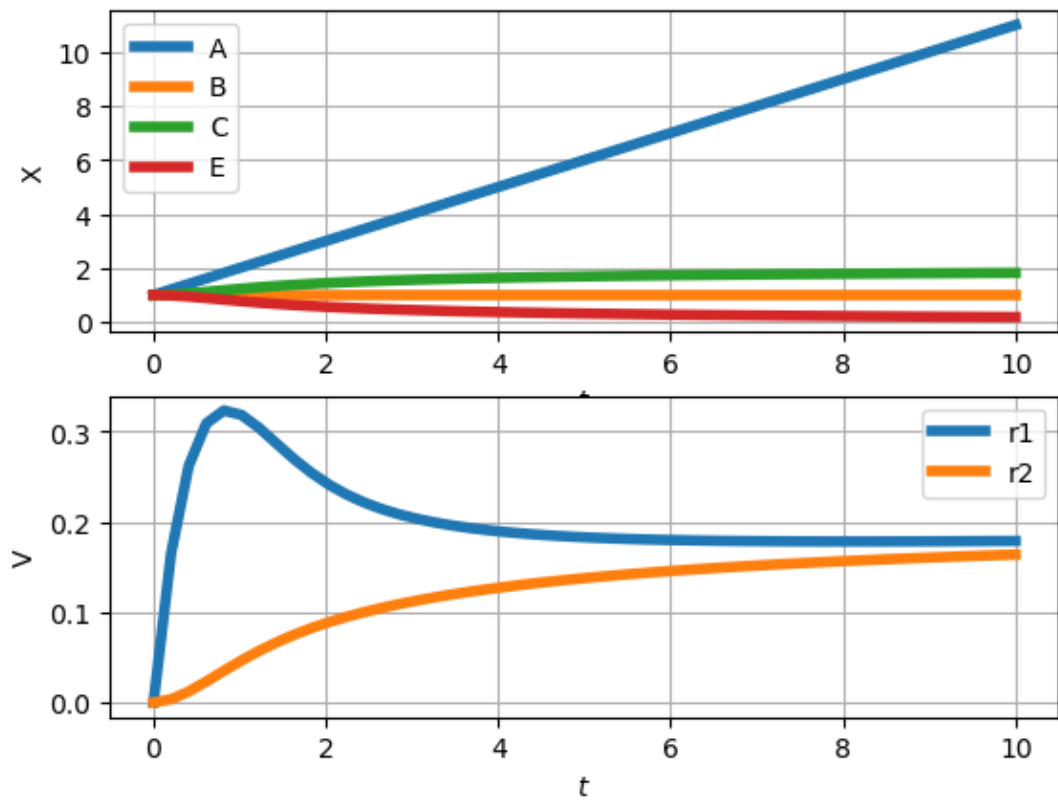


As the product B amount is greater than that of the substrate A , the flow proceeds in reverse.

2.5.1 Time-varying chemostats

By default, chemostats remain at the corresponding initial state. This can be changed by declaring a time-varying expression for the chemostat state. For example, set the chemostat for substrate A to have a value of $1 + t$:

```
[24]: X_chemo = {'A': '1+t'}
      result = st.sim(s,t=t,sc=sc,parameter=parameter,X_chemo=X_chemo)
      st.plot(s,result)
```



Note that the flow rates reach a maximum value as the amount of enzyme x_E reduces to zero. This behaviour is typical of systems with conserved moities in general and enzyme catalysed reactions in particular.

References