

Dynamical modeling of reaction networksMathematics of system biology

Romain Yvinec

► To cite this version:

Romain Yvinec. Dynamical modeling of reaction networksMathematics of system biology. 2020 Workshop - 'bioinformatics and biomathematical approaches to integrate the GPCR signal, Nov 2020, Online, France. pp.1-70. hal-03115099

HAL Id: hal-03115099 https://hal.science/hal-03115099v1

Submitted on 19 Jan2021

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Dynamical modeling of reaction networks Mathematics of system biology

Romain Yvinec

Systems Biology and Networks

Motivations and Objectives

Chemical reaction network formalism

Parameter estimation

Applications

Systems Biology and Networks

Motivations and Objectives

Chemical reaction network formalism

Parameter estimation

Applications



Small networks in Population dynamics

(Interactions between populations, Epidemiology)

Lotka-Volterra model

$$\begin{array}{ccc} \varnothing & \stackrel{k_1}{\longrightarrow} & A \\ A + B & \stackrel{k_2}{\longrightarrow} & 2B \\ B & \stackrel{k_3}{\longrightarrow} & \varnothing \end{array}$$





Small networks in molecular biology ('Toy' molecular models with isolated components)



Pharmacology model $R_i \rightleftharpoons R_a$

 $AR_2 \simeq AR_i$

 $\begin{array}{rccc} A + R_i & \rightleftharpoons & AR_i \\ A + R_a & \rightleftharpoons & AR_a \end{array}$

(Single) Gene Expression

$$\begin{array}{cccc} G & \stackrel{\lambda_1}{\longrightarrow} & G + M \\ M & \stackrel{\lambda_2}{\longrightarrow} & M + P \\ M & \stackrel{\gamma_1}{\longrightarrow} & \varnothing \\ P & \stackrel{\gamma_2}{\longrightarrow} & \varnothing \\ G & \stackrel{k_{off}}{\overleftarrow{k_{on}}} & G_{off} \end{array}$$

Eldar and Elowitz (Nature 2010)



Co-expression genes network









Systems Biology and Networks

Motivations and Objectives

Chemical reaction network formalism

Parameter estimation

Applications

- Understand non-trivial behavior of a biological system (by reproducing this behavior with an understandable model, starting from 'first principles')
- Help to identify key regulatory process in signaling cascades
- Quantify some non-observables quantities, in particular : molecules concentrations, reaction rates.

Possible applications of mathematical modelling

Today : Understand the mathematical formalism of dynamical reactions network

- **Today :** Understand the mathematical formalism of dynamical reactions network
 - Build a model from a network of interactions [Cell Designer]
 - Parameter calibration with kinetic data [GraphPad Prism / Copasi]

Warning !



#Ihatemathematics

The following couples of slides contain some abstract notions... Why that $? \end{tabular}$

Maths is about structures...and is generic !



Reactions networks applications range from

- Chemistry
- molecular biology
- epidemiology
- and beyond !





Systems Biology and Networks

Motivations and Objectives

Chemical reaction network formalism

Parameter estimation

Applications

Definition

A chemical reaction network is given by thee sets $(\mathcal{S},\mathcal{C},\mathcal{R})$:

- Species, S := {S₁, · · · , S_d} : molecules that undergo a serie of chemical reactions.
- Reactant / Product, C := {y¹, · · · yⁿ} : Linear combination of species, that represent either 'what is consumed', or 'what is produced', in any reaction.
- Reaction, R := {y^k → y^{k'}, y^k, y^{k'} ∈ C} : ensemble of reactions between species or combination of species (directed graph between Reactant / Product).

Definition

A chemical reaction network is given by thee sets $(\mathcal{S},\mathcal{C},\mathcal{R})$:

- Species, S := {S₁, · · · , S_d} : molecules that undergo a serie of chemical reactions.
- Reactant / Product, C := {y¹, · · · yⁿ} : Linear combination of species, that represent either 'what is consumed', or 'what is produced', in any reaction.
- Reaction, R := {y^k → y^{k'}, y^k, y^{k'} ∈ C} : ensemble of reactions between species or combination of species (directed graph between Reactant / Product).
- Mass-action law, κ : a list of positive parameter (kinetic rate) for each reaction in $\mathcal R$

Example $A \stackrel{k^+}{\Longrightarrow} B$ $\begin{array}{l} \text{Species } \mathcal{E} := \{A, B\} \\ \text{R} \ / \ \text{P} \ \mathcal{C} := \{A, B\} \\ \text{Reaction } \mathcal{R} := \{A \rightarrow B, B \rightarrow A\} \end{array}$ Rate $\{k^+, k^-\}$

Chemical Reaction Network, vocabulary

Example (minimal cAMP production model)

$$L + R \xrightarrow[k_{off}]{k_{off}} LR$$

$$ATP + LR \xrightarrow[k^+]{k^-} CAMP + LR$$

$$CAMP \xrightarrow[k^-]{k^-} AMP$$

Species
$$\mathcal{E} := \{L, R, LR, ATP, cAMP, AMP\}$$

 $R / P \mathcal{C} := \{L + R, LR, ATP + LR, cAMP + LR, cAMP, AMP\}$
Reaction $\mathcal{R} := \{L + R \rightarrow LR, LR \rightarrow L + R, ATP + LR \rightarrow cAMP + LR, cAMP \rightarrow AMP\}$
Rate $\{k_{on}, k_{off}, k^+, k^-\}$

Chemical Reaction Network, FSHR-induced cAMP signals



Figure – FSHR (G α s-coupled GPCR) models build on CellDesigner

Cell Designer



...with well-defined conventions

About diagrams and layout...

Funahashi, A., Tanimura, N., Morohashi, M., and Kitano, H., CellDesigner : a process diagram editor for gene-regulatory and biochemical networks, BIOSILICO, 1 :159-162, 2003

Chemical Reaction Network, "real" example



Figure – ERK Phosphorylation pathways build on CellDesigner, Heitzler et al. MSB 2012

- Many examples of dynamical system biology models
- The formalism of *reaction network* models
- How to build a reaction network within Cell Designer

- Many examples of dynamical system biology models
- The formalism of reaction network models
- How to build a reaction network within Cell Designer

NB : there exists public databases of reaction network models.



- Many examples of dynamical system biology models
- The formalism of *reaction network* models
- How to build a reaction network within Cell Designer
- NB : there exists public databases of reaction network models.



NB (bis) : A reaction network is a network... but a network is **NOT** a reaction network !

- Many examples of dynamical system biology models
- The formalism of reaction network models
- How to build a reaction network within Cell Designer
- What about dynamics?

Warning !



#Ihatemathematics

The following couples of slides contain some abstract notions...let's try to catch the meaning !

We build a model that

- Keep track of concentration of species along time.
- Satisfy Law of Mass action : The velocity of a reaction is proportional to the concentrations of its reactants.
- Is a system of Ordinary Differential Equations, in which reactions are "added" on top of each other, e.g. they happens continuously and simultaneously.

Example $A \stackrel{2}{\underset{0.1}{\longrightarrow}} B$ $\frac{dx_A}{dt}(t) = -0.1x_A(t) + 2x_B(t), \quad x_A(t=0) = A_{tot}$ $\frac{dx_B}{dt} = +0.1x_A(t) - 2x_B(t), \quad x_B(t=0) = 0$ $x_A(t) = \text{time dependent concentration of species } A$

Example

$$L + R \stackrel{k_{on}}{\underset{k_{off}}{\longleftarrow}} LR$$

$$\frac{dx_L}{dt}(t) = -k_{on}x_L(t)x_R(t) + k_{off}x_{LR}(t), \quad x_L(0) = Dose \frac{dx_R}{dt}(t) = -k_{on}x_L(t)x_R(t) + k_{off}x_{LR}(t), \quad x_R(0) = R_{tot} \frac{dx_{LR}}{dt}(t) = k_{on}x_L(t)x_R(t) - k_{off}x_{LR}(t), \quad x_{LR}(0) = 0.$$

Chemical Reaction Network and Dynamical models

Example (minimal cAMP production model)

$$\begin{array}{ccc} L+R & \stackrel{k_{on}}{\longrightarrow} & LR \\ ATP+LR & \stackrel{k^+}{\longrightarrow} & cAMP+LR \\ cAMP & \stackrel{k^-}{\longrightarrow} & AMP \end{array}$$

$$\begin{aligned} \frac{dx_L}{dt} &= -k_{on}x_Lx_R + k_{off}x_{LR}, \quad x_L(0) = Dose \\ \frac{dx_R}{dt} &= -k_{on}x_Lx_R + k_{off}x_{LR}, \quad x_R(0) = R_{tot} \\ \frac{dx_{LR}}{dt} &= k_{on}x_Lx_R - k_{off}x_{LR}, \quad x_{LR}(0) = 0 \\ \frac{dx_{cAMP}}{dt} &= k^+x_{ATP}x_{LR} - k^-x_{cAMP}, \quad x_{cAMP}(0) = 0 \\ \frac{dx_{ATP}}{dt} &= -k^+x_{ATP}x_{LR} \quad x_{ATP}(0) = ATP_{tot}. \end{aligned}$$

But what is an "Ordinary Differential Equation"? A math theory in one slide !

The equation

$$\frac{dx}{dt}=v(x)\,,$$

is numerically solved by successive time-step iteration, of small length $\Delta t \ll 1$:

1) Start at a given initial condition x_0 at time $t_0 = 0$

But what is an "Ordinary Differential Equation"? A math theory in one slide !

The equation

$$\frac{dx}{dt} = v(x) \,,$$

is numerically solved by successive time-step iteration, of small length $\Delta t \ll 1$:

- 1) Start at a given initial condition x_0 at time $t_0 = 0$
- 2) To calculate the value of x at the first time step, remember that (assuming constant speed)

Final Position = Initial Position + velocity * Time,

which becomes, in mathematical notations,

$$x(\Delta t) = x_0 + v(x_0) * \Delta t \,,$$

But what is an "Ordinary Differential Equation"? A math theory in one slide!

The equation

$$\frac{dx}{dt} = v(x) \,,$$

is numerically solved by successive time-step iteration, of small length $\Delta t \ll 1$:

- 1) Start at a given initial condition x_0 at time $t_0 = 0$
- 2) To calculate the value of x at the first time step, remember that (assuming constant speed)

Final Position = Initial Position + velocity * Time,

which becomes, in mathematical notations,

$$x(\Delta t) = x_0 + v(x_0) * \Delta t \, ,$$

Iterate : To calculate the value of x at the next time step, use

$$x((i+1)*\Delta t) = x(i*\Delta t) + v(x(i*\Delta t))*\Delta t,$$

But what is an "Ordinary Differential Equation"? A math theory in one slide...and a figure !



Figure – Solving an ODE

But what is an "Ordinary Differential Equation"? A math theory in one slide...and a figure !



Figure – Solving an ODE

But what is an "Ordinary Differential Equation"? A math theory in one slide...and a figure !



Figure – Solving an ODE



ODE solver within Cell Designer

Funahashi, A., Tanimura, N., Morohashi, M., and Kitano, H., CellDesigner : a process diagram editor for gene-regulatory and biochemical networks, BIOSILICO, 1 :159-162, 2003

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
- How to *build and simulate* a reaction network within Cell Designer

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
- How to *build and simulate* a reaction network within Cell Designer

 NB : You don't need to code, but you need to specify kinetic rate and initial condition values to simulate a reaction network.

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
- How to *build and simulate* a reaction network within Cell Designer

 NB : You don't need to code, but you need to specify kinetic rate and initial condition values to simulate a reaction network.

NB (bis) : You can play with this tools to "explore" the behavior of a model. But that can be time consuming and inefficient...

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
- How to *build and simulate* a reaction network within Cell Designer

What about inferring those values from data?

Systems Biology and Networks

Motivations and Objectives

Chemical reaction network formalism

Parameter estimation

Applications

Parameter and network inference in Chemical Reaction Network

Goal : Given some time series data, find the minimal (biologically plausible) reaction network with its parameter (=reaction rates and initial conditions) that fits consistently the data.

Parameter and network inference in Chemical Reaction Network

Goal : Given some time series data, find the minimal (biologically plausible) reaction network with its parameter (=reaction rates and initial conditions) that fits consistently the data.



Parameter and network inference in Chemical Reaction Network

Goal : Given some time series data, find the minimal (biologically plausible) reaction network with its parameter (=reaction rates and initial conditions) that fits consistently the data.



Regression analysis and Parameter estimation with time series : What is difficult?

- In linear models, there exists a unique optimal solution
- Yet in practice, (generalized) linear models do not perform well on biochemical data due to Heteroscedasticity and highly dependent time point data.

Regression analysis and Parameter estimation with time series : What is difficult ?

- For most of (nonlinear) reaction network, there is no guarantee to find a unique optimal solution.
- Reaction network models allows to perform multifactorial analysis ("Anova-like")

Regression analysis and Parameter estimation with time series : What is difficult ?

- For most of (nonlinear) reaction network, there is no guarantee to find a unique optimal solution.
- Reaction network models allows to perform multifactorial analysis ("Anova-like")
- Many other tools exists from the statistical field of *time series* analysis.

Warning !



#Ihatemathematics

The following couples of slides contain some abstract notions...but that the last ones !

- **Goal :** Given some time series data, find the minimal (biologically plausible) reaction network with its parameter (=reaction rates and initial conditions) that fits consistently the data.
- **Strategy** 1) From a given network (S, C, R), with given parameter values, solve the ODEs,

$$\frac{dx}{dt}=v(x,k)\,,\quad x(0)=x_0\,,$$

and compute a distance between the solution and the data.

- **Goal :** Given some time series data, find the minimal (biologically plausible) reaction network with its parameter (=reaction rates and initial conditions) that fits consistently the data.
- **Strategy** 1) From a given network (S, C, R), with given parameter values, solve the ODEs,

$$\frac{dx}{dt}=v(x,k)\,,\quad x(0)=x_0\,,$$

and compute a distance between the solution and the data.
Strategy 2) Using optimization algorithms, find the best parameter values k, x₀, to minimize the distance

- **Goal** : Given some time series data, find the minimal (biologically plausible) reaction network with its parameter (=reaction rates and initial conditions) that fits consistently the data.
- **Strategy** 1) From a given network (S, C, R), with given parameter values, solve the ODEs,

$$\frac{dx}{dt}=v(x,k)\,,\quad x(0)=x_0\,,$$

and compute a distance between the solution and the data.

- **Strategy** 2) Using **optimization** algorithms, find the best parameter values k, x_0 , to minimize the distance
- **Strategy** 3) If needed, change the reaction network (add or delete species/reactions)

- **Goal :** Given some time series data, find the minimal (biologically plausible) reaction network with its parameter (=reaction rates and initial conditions) that fits consistently the data.
- **Statistics** There exists a well developed statistical theory to assess the **quality of a fit** and to give **confidence interval** on parameter values (-> See Likelihood maximization or Bayesian statistics).

Parameter estimation in



Predefined or user-defined time-dependent equations

Model

Radioligand=HotNM*1e-9 Kob=[Radioligand]*Kon+Koff Kd=Koff/Kon Eq=Bmax*radioligand/(radioligand + Kd) Association=Eq*(1-exp(-1*Kob*Time0)) YatTime0 = Eq*(1-exp(-1*Kob*Time0)) Dissociation=YatTime0*exp(-1*Kof*(X-Time0)) Y=IF(X<Time0, Association, Dissociation) + NS



Parameter estimation in



- Limited to solvable models
- Adapted to analyze one single output at a time, assuming excess of Ligand.



Predefined or user-defined time-dependent equations



Hoare et al., Analyzing kinetic signaling data for G-protein-coupled receptors, Scientific Reports 10(1) :12263 2020

Parameter estimation in Copasi



- Models can be imported from Cell Designer.
- Supports both graphical interface and command line.

Bergman et al. COPASI and its applications in biotechnology, Journal of Biotechnology 261 :215-220, 2017. Hoops et al. COPASI : a COmplex PAthway SImulator. Bioinformatics

22:3067-74, 2006.

Remember ! it's an iterative and interdisciplinary workflow !



Remember ! it's an iterative and interdisciplinary workflow !





- There is a trade-off between toy minimal models and detailed biochemistry pathways.
- Overfitting leads to unreliable prediction and meaningless model / parameter value.
- (Advanced) statistical tools exist to sort this out : model selection (especially for hierarchical models) and parameter identifiability.

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
- How to build and simulate a reaction network model within Cell Designer.
- How to calibrate parameters of a dynamical reaction network model with GraphPad Prism and/or Copasi.

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
- How to build and simulate a reaction network model within Cell Designer.
- How to calibrate parameters of a dynamical reaction network model with GraphPad Prism and/or Copasi.

 ${\sf NB}$: The full workflow can be long and require collaboration with statistician / applied mathematician.

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
- How to build and simulate a reaction network model within Cell Designer.
- How to calibrate parameters of a dynamical reaction network model with GraphPad Prism and/or Copasi.

NB bis : What about applications? Go to session 8 : Computational approaches!

Systems Biology and Networks

Motivations and Objectives

Chemical reaction network formalism

Parameter estimation

Applications

Some applications : Understanding G protein activation cycle







Bridge, Meads, Frattini, Winfield, Ladds, Modelling and simulation of biased agonism dynamics at a G protein-coupled receptor, J. Theoret. Biol. 442 :44–65, 2018

Some applications : Shedding light on GPCR-induced Calcium oscillations in Astrocytes



De Pittà, Ben-Jacob, Berry, G protein-coupled receptor-mediated calcium signaling in astrocytes, in Computational Glioscience, Springer 2019.

Some applications : Revisiting signaling bias using dynamical model



- Dynamical reaction network framework has many different applications.
- Its a powerful framework to reveal comprehensive spatio-temporal patterns behind GPCR signaling complexities.
- Its a powerful framework to analyze quantitatively time series data in GPCR signaling.
- Adequate tools foster necessary interdisciplinary collaborations by providing a common language.