

Dynamical modeling of reaction networks. Application to biased signaling

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Submitted on 19 Jul 2022

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Dynamical modeling of reaction networks Application to biased signaling

Romain Yvinec

BIOS team, Physiologie de la Reproduction et des Comportements, INRAE Tours, MUSCA team, INRIA Saclay,

France

Systems Biology and Networks

Reaction network formalism

Viewing reaction network and Databases

Dynamics model with reaction network

Parameter estimation

Applications

Outline

Systems Biology and Networks

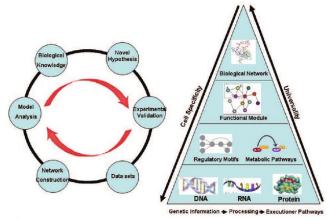
Reaction network formalism

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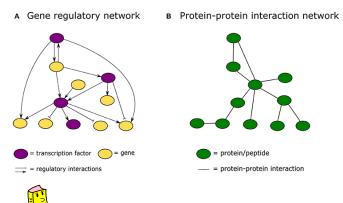
Parameter estimation

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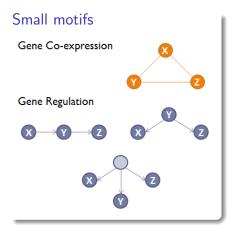


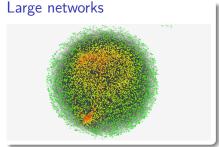


Oltvai and Barabasi, Science 25:763-764, 2002.

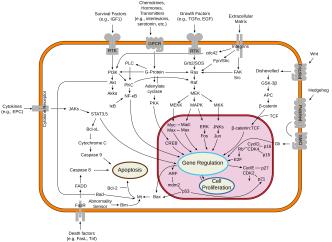


Vandereyken et al., Front. Plant Sci., 2018.





Signaling pathways

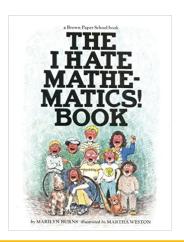


Network, graph and reaction network

Foday: We will deal with a specific kind of network, that have a higher degree of structure:

- ▶ Petri Nets, Species-Reaction graph, Reaction Network...
- It contains entities (proteins, genes, metabolites...) and processes that modifies the entities abundances through time

Warning!



#Ihatemathematics

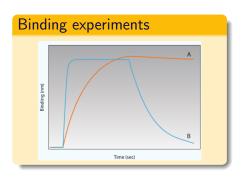
The following couples of slides contain some abstract notions... Why that?

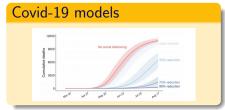
Maths is about structures...and is generic!



Reactions networks applications range from

- Chemistry
- molecular biology
- epidemiology
- ▶ and beyond!





Outline

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Reaction Network, vocabulary

Definition

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

- Species, $S := \{S_1, \dots, S_d\}$: molecules that undergo a serie of chemical reactions.
- Reactant / Product, $C := \{y^1, \dots y^n\}$: Linear combination of species, that represent either 'what is consumed', or 'what is produced', in any reaction.
- Reaction, $\mathcal{R} := \{y^k \to y^{k'}, y^k, y^{k'} \in \mathcal{C}\}$: ensemble of reactions between species or combination of species (directed graph between Reactant / Product).

Reaction Network, vocabulary

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- Mass-action law, κ : a list of positive parameter (kinetic rate) for each reaction in $\mathcal R$

Chemical Reaction Network, vocabulary

Example

$$A \stackrel{k^+}{\rightleftharpoons} B$$

Species
$$\mathcal{E} := \{A, B\}$$

R / P $\mathcal{C} := \{A, B\}$
Reaction $\mathcal{R} := \{A \to B, B \to A\}$
Rate $\{k^+, k^-\}$

Reaction Network, vocabulary

Example (small cAMP production model)

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

$$ATP + LR \xrightarrow{k^{+}} cAMP + LR$$

$$cAMP \xrightarrow{k^{-}} AMP$$

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

$$R / P \mathcal{C} := \{L + R, LR, ATP + LR, cAMP + LR, cAMP, AMP\}$$

$$Reaction \mathcal{R} := \{L + R \rightarrow LR, LR \rightarrow L + R, ATP + LR \rightarrow cAMP + LR, cAMP \rightarrow AMP\}$$

$$Rate \{k_{on}, k_{off}, k^{+}, k^{-}\}$$

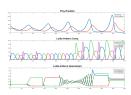
Examples of Reaction Networks at different scales

Small networks in Population dynamics

(Interactions between populations, Epidemiology)

Lotka-Volterra model

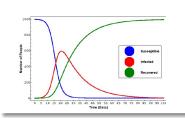
$$\begin{array}{ccc}
\varnothing & \xrightarrow{k_1} & A \\
A+B & \xrightarrow{k_2} & 2B \\
B & \xrightarrow{k_3} & \varnothing
\end{array}$$



S.I.R model

$$S + I \xrightarrow{k_1} I + I$$

$$I \xrightarrow{k_2} R$$



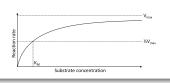
Examples of Reaction Networks at different scales

Small networks in molecular biology

('Toy' molecular models with isolated components)

Enzymatic kinetics

$$E + S \xrightarrow[k_1^-]{k_1^+} ES \xrightarrow[k_2^-]{k_2^+} E + P$$



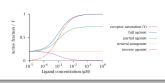
Pharmacology model

$$R_{i} \rightleftharpoons R_{a}$$

$$A + R_{i} \rightleftharpoons AR_{i}$$

$$A + R_{a} \rightleftharpoons AR_{a}$$

$$AR_{a} \rightleftharpoons AR_{i}$$



Examples of Reaction Networks at different scales

(Single) Gene Expression

$$G \xrightarrow{\lambda_{1}} G + M$$

$$M \xrightarrow{\lambda_{2}} M + P$$

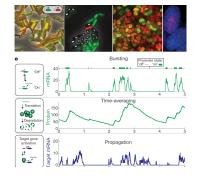
$$M \xrightarrow{\gamma_{1}} \varnothing$$

$$P \xrightarrow{\gamma_{2}} \varnothing$$

$$G \xrightarrow{k_{off}} G_{off}$$



Eldar and Elowitz (Nature 2010)



Outline

Systems Biology and Networks

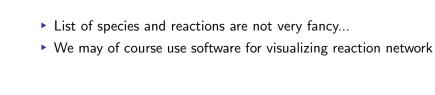
Reaction network formalism

Viewing reaction network and Databases

Dynamics model with reaction network

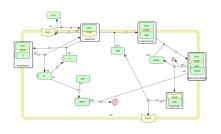
Parameter estimation

Applications



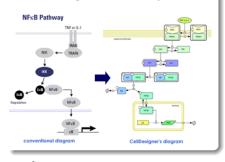
Reaction Network Visualization

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

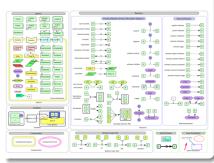


Cell Designer

About diagrams and layout...



...with well-defined conventions



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

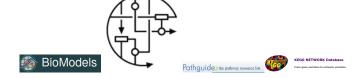
We have seen

- Examples of network and reaction network models
- ▶ The formalism of *reaction network* models
- ▶ How to build a reaction network within Cell Designer

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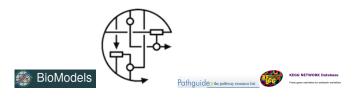
NB: there exists public databases of (reaction) network models.



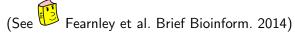
We have seen

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- ▶ The formalism of *reaction network* models
- ▶ How to *build* a reaction network within Cell Designer

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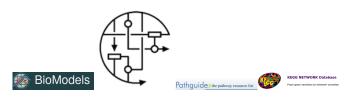
But! Extracting reaction network from database require some care



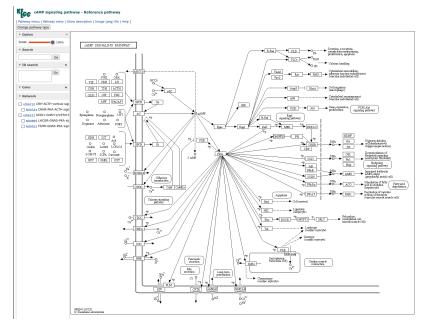
We have seen

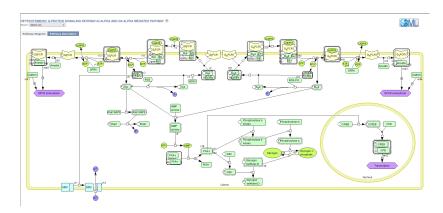
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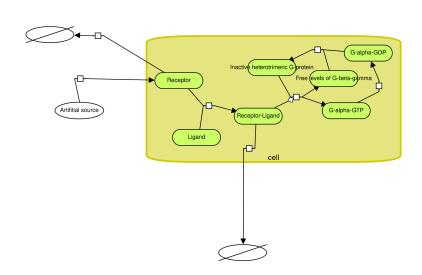
NB : there exists public databases of (reaction) network models.



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







We have seen

- Examples of network and reaction network models
- ▶ The formalism of reaction network models
- ▶ How to build a reaction network within Cell Designer
- What about dynamics?

Outline

Systems Biology and Networks

Reaction network formalism

Viewing reaction network and Database

Dynamics model with reaction network

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Possible applications of Dynamics modeling

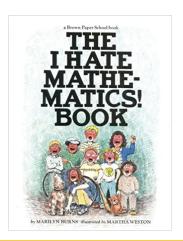
- Understand non-trivial behavior of a biological system (by reproducing this behavior with an understandable model, starting from 'first principles')
- ▶ Help to identify key regulatory process in signaling cascades
- Quantify some non-observables quantities, in particular : molecules concentrations,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

Possible applications of Dynamics modeling

Now : Understand the mathematical formalism of dynamical reactions network

- Build and simulate a dynamical model from a reaction network [Cell Designer]
- Parameter calibration with kinetic data [GraphPad Prism / Copasi]

Warning!



#Ihatemathematics

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

Chemical Reaction Network and Dynamical models

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 (Rate equations).

Chemical Reaction Network and Dynamical models

Example

$$A \stackrel{2}{\rightleftharpoons} B$$

$$\begin{array}{lcl} \frac{dx_A}{dt}(t) & = & -0.1x_A(t) + 2x_B(t) \,, & x_A(t=0) = A_{tot} \\ \frac{dx_B}{dt} & = & +0.1x_A(t) - 2x_B(t) \,, & x_B(t=0) = 0 \end{array}$$

 $x_A(t)$ = time dependent concentration of species A

Chemical Reaction Network and Dynamical models

Example

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

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

$$\frac{kLR}{dt}(t) = k_{on}x_L(t)x_R(t) - k_{off}x_{LR}(t), \quad x_{LR}(0) = 0.$$

Chemical Reaction Network and Dynamical models

Example (minimal cAMP production model)

$$L + R \xrightarrow{k_{on}} LR$$

$$ATP + LR \xrightarrow{k^{+}} cAMP + LR$$

$$cAMP \xrightarrow{k^{-}} AMP$$

$$\frac{dx_{L}}{dt} = -k_{on}x_{L}x_{R} + k_{off}x_{LR}, \quad x_{L}(0) = Dose$$

$$\frac{dx_{R}}{dt} = -k_{on}x_{L}x_{R} + k_{off}x_{LR}, \quad x_{R}(0) = R_{tot}$$

$$\frac{dx_{LR}}{dt} = k_{on}x_{L}x_{R} - k_{off}x_{LR}, \quad x_{LR}(0) = 0$$

$$\frac{dx_{CAMP}}{dt} = k^{+}x_{ATP}x_{LR} - k^{-}x_{CAMP}, \quad x_{CAMP}(0) = 0$$

$$\frac{dx_{CAMP}}{dt} = -k^{+}x_{ATP}x_{LR} \quad x_{ATP}(0) = ATP_{tot}.$$

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 $t_0 = 0$

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- 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,

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

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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!

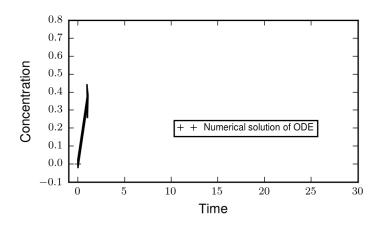


Figure - Solving an ODE

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

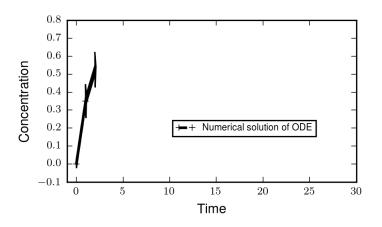


Figure – Solving an ODE

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

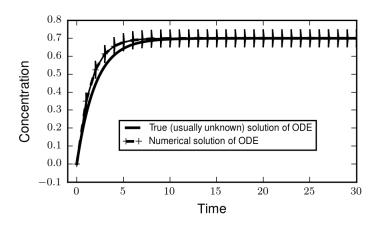
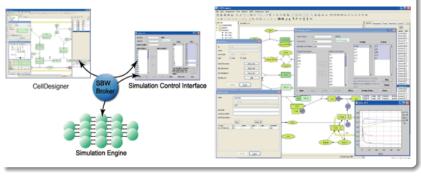


Figure - Solving an ODE

Solving an ODE in practice : no need to code!

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

We have seen

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

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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...

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What about inferring those values from data?

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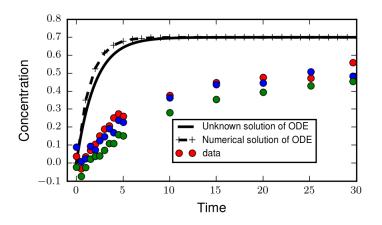
Applications

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.

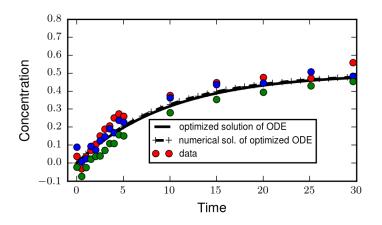
Parameter and network inference in Chemical Reaction Network

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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.

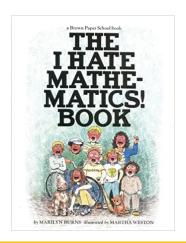
Regression analysis and Parameter estimation with time series : What is difficult?

- ► 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.

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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

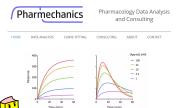
Radioligand=HotNM*1e-9
Kob=[Radioligand]*Kon+Koff
Kd=Koff/Kon
Eq=Bmax*radioligand/(radioligand + Kd)
Association=Eq*(1-exp(-1*Kob*X))
YatTime0 = Eq*(1-exp(-1*Kob*Time0))
Dissociation= YatTime0*exp(-1*Koff*(X-Time0))
Y=IF(X<Time0. Association. Dissociation) + NS



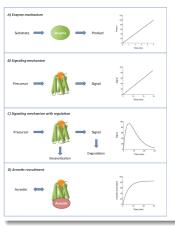
Parameter estimation in



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

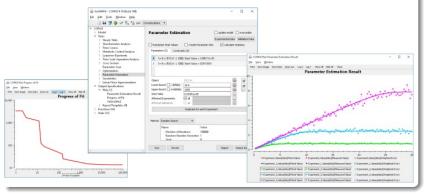


Predefined or user-defined time-dependent equations



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

Parameter estimation in Copasi

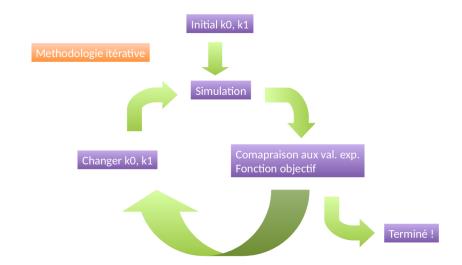


- Models can be imported from Cell Designer.
- Supports both graphical interface and command line.

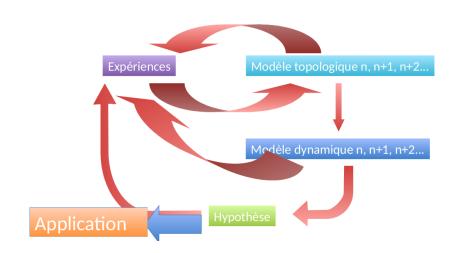
Bergman et al. COPASI and its applications in biotechnology, Journal of Biotechnology 261:215-220, 2017.

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!



Is the monkey who typed Hamlet actually a good writer?

Overfitting

- 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.

We have seen

- 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.

We have seen

- Many examples of dynamical system biology models
- The formalism of dynamical reaction network models
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NB : The full workflow can be long and require collaboration with statistician / applied mathematician.

We have seen

- 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?

Outline

Systems Biology and Networks

Reaction network formalism

Viewing reaction network and Database

Dynamics model with reaction network

Parameter estimation

Applications

Motivations and Case study

Use reaction network modeling (kinetic pathway) to

- Fully exploit kinetic data
- Give more mechanistic insight of signaling bias
- Develop a parsimonious and statistically significant framework to characterize pharmacological ligand properties

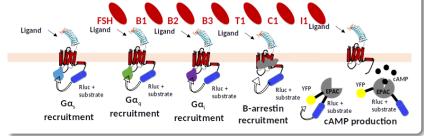
Motivations and Case study

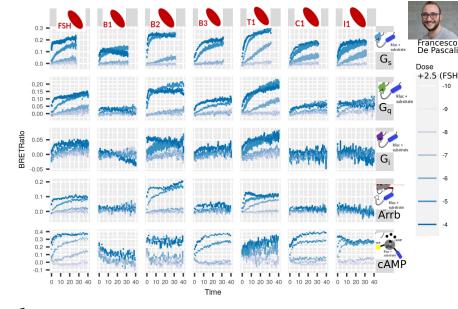
Use reaction network modeling (kinetic pathway) to

- Fully exploit kinetic data
- Give more mechanistic insight of signaling bias
- Develop a parsimonious and statistically significant framework to characterize pharmacological ligand properties

Case study on FSHR

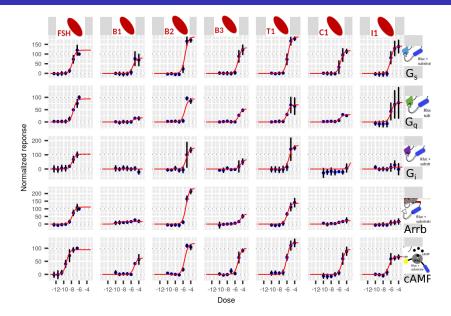
- \star 5 BRET sensors : NES-Venus mG, yPET- β -arrestin 2, Camyel
- * FSH + 6 LMW compounds (Benzamides, Thiazolidinone, Chromenopyrazole, Imidazole) (TocopheRx, Burlington, VT, USA).



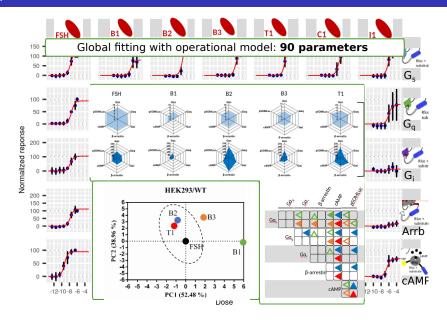


De Pascali et al. Pharmacological Characterization of Low Molecular Weight Biased Agonists at the Follicle Stimulating Hormone Receptor, IJMS, 2021.

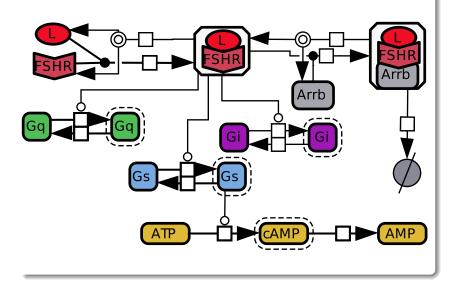
Operational model with A.U.C



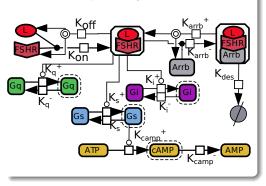
Operational model with A.U.C



Reaction network: multiple Pathways modeling

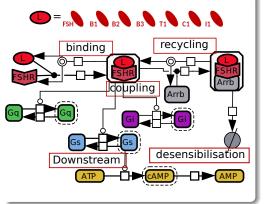


Generate all pathways at once



- Dynamic reaction networks (ODE) keep track of concentration of each molecule along time.
- Parameters: initial quantity of molecules and kinetic rates (13).

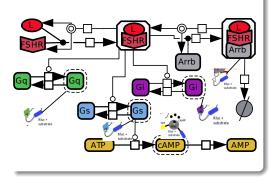
Mechanistic link with data



We hypothesize that

- Kinetic rate values reflects pharmacological ligand properties.
- Measurements are performed in a same cellular context.
- Measurements are proportional to concentration of molecules.

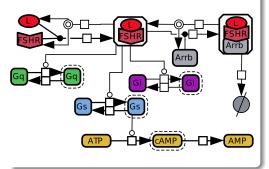
Mechanistic link with data

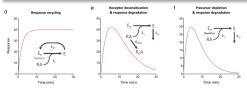


We hypothesize that

- Kinetic rate values reflects pharmacological ligand properties.
- Measurements are performed in a same cellular context.
- Measurements are proportional to concentration of molecules.

Signaling profile diversity





- The model is "minimal" (model selection criteria)
- We generalize recent attempts to define a "kinetic operational model" (Watch Nicola Dijon's flash presentation)

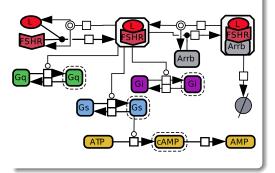


Hoare et al., Analyzing kinetic

signaling data for G-protein-coupled receptors,

Scientific Reports 2020

Global fitting enforcing sparsity



- Our method is a global fitting approach (all pathways, all ligand).
- We enforce Ligand specific parameters through penalization.



Raue et al., Data2Dynamics : a

modeling environment tailored to parameter estimation in dynamical systems,

Bioinformatics 2015

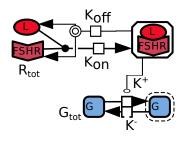


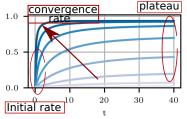
Steiert et al., L1 regularization

facilitates detection of cell type-specific parameters in dynamical systems,

Bioinformatics 2016

Can we really infer parameter from kinetic data?





Initial rate

$$\frac{1}{2}R_{tot}G_{tot}k_{on}k^{+}[L]t^{2}$$

Equilibrium

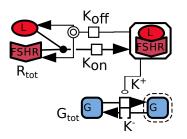
$$\frac{R_{tot}G_{tot}[L]}{K_A + (R_{tot} + K_E)[L]}$$

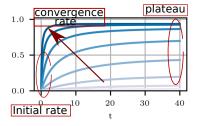
Convergence rate

$$k^{+}\frac{R_{tot}[L]}{K_{A}+[L]}+k^{-}$$

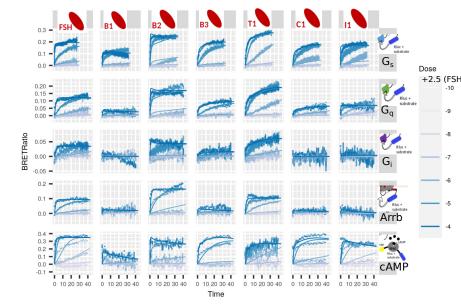
$$K_A = \frac{K_{off}}{K_{on}}$$
 , $K_E = \frac{K^-}{K^+}$

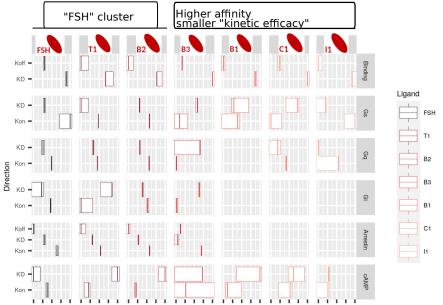
Can we really infer parameter from kinetic data?



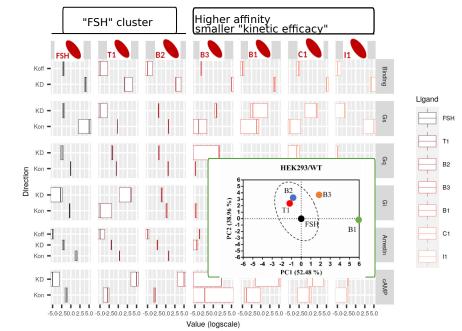


- Initial rate $\frac{1}{2}R_{tot}G_{tot}k_{on}k^{+}[L]t^{2}$
- Equilibrium $\frac{R_{tot}G_{tot}[L]}{K_A + (R_{tot} + K_E)[L]}$
- Convergence rate $k^+ \frac{R_{tot}[L]}{K_A + [L]} + k^-$
- → In practice the global fitting improves parameter identifiability.
- → Low doses and long time signal are important.





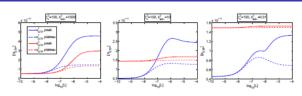
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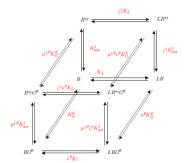
Conclusions

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

Some applications : Understanding G protein activation cycle

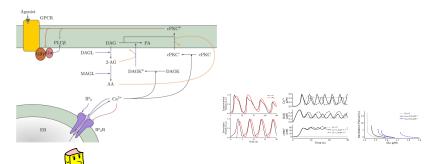






Bridge, Meads, Frattini, Winfield, Ladds, Modelling and simulation of biased agonism dynamics at a G protein-coupled receptor, J. Theoret. Biol. 442:44–65, 2018

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.