

Computing dynamics of (gene) regulatory networks

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Outline

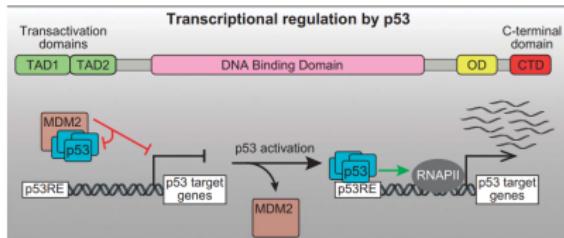
1 Motivation

2 Framework

- Step 1: Combinatorial model
- Step 2: Combinatorial dynamics
- Step 3: Continuous dynamics
- Step 4: Validation

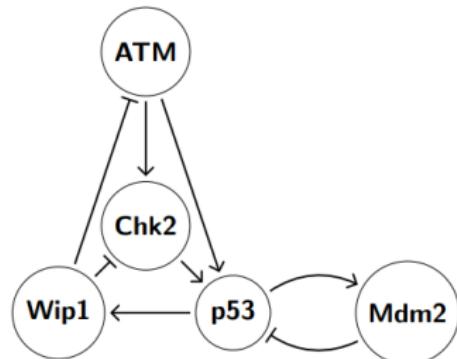
3 Applications

Gene regulation



Graphical Abstract

(a) p53

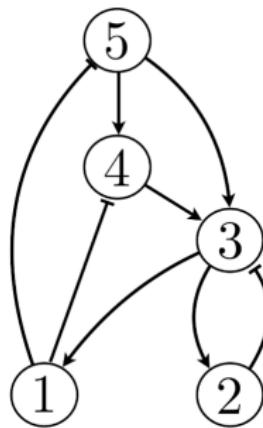


(b) Subnetwork of key species of the p53 signaling network

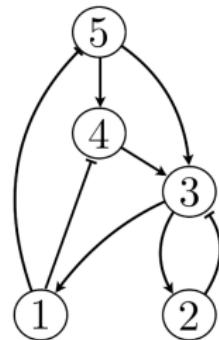
(Gene) Regulatory Networks

Definition

A regulatory network is a directed graph $G = (V, E, W)$ with a set of vertices V , edges E and a sign function $W : E \rightarrow \{+1, -1\}$.

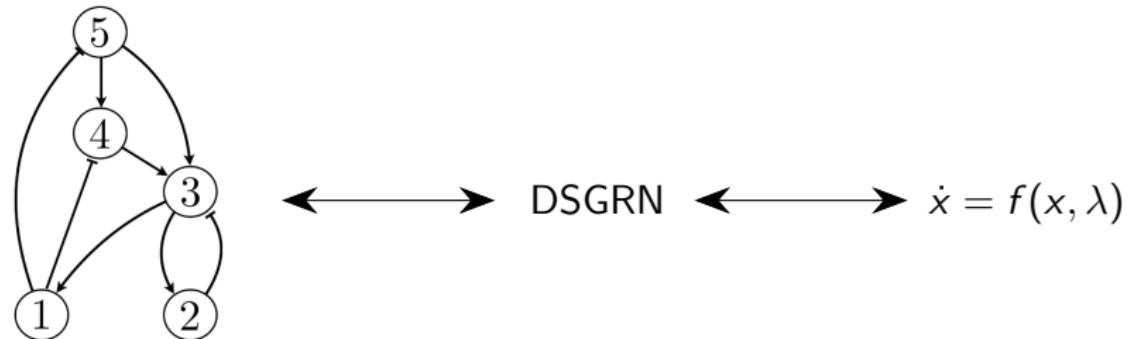


Goal



$$\dot{x} = f(x, \lambda)$$

Goal



Dynamic Signatures Generated by Regulatory Networks (DSGRN)

Main steps

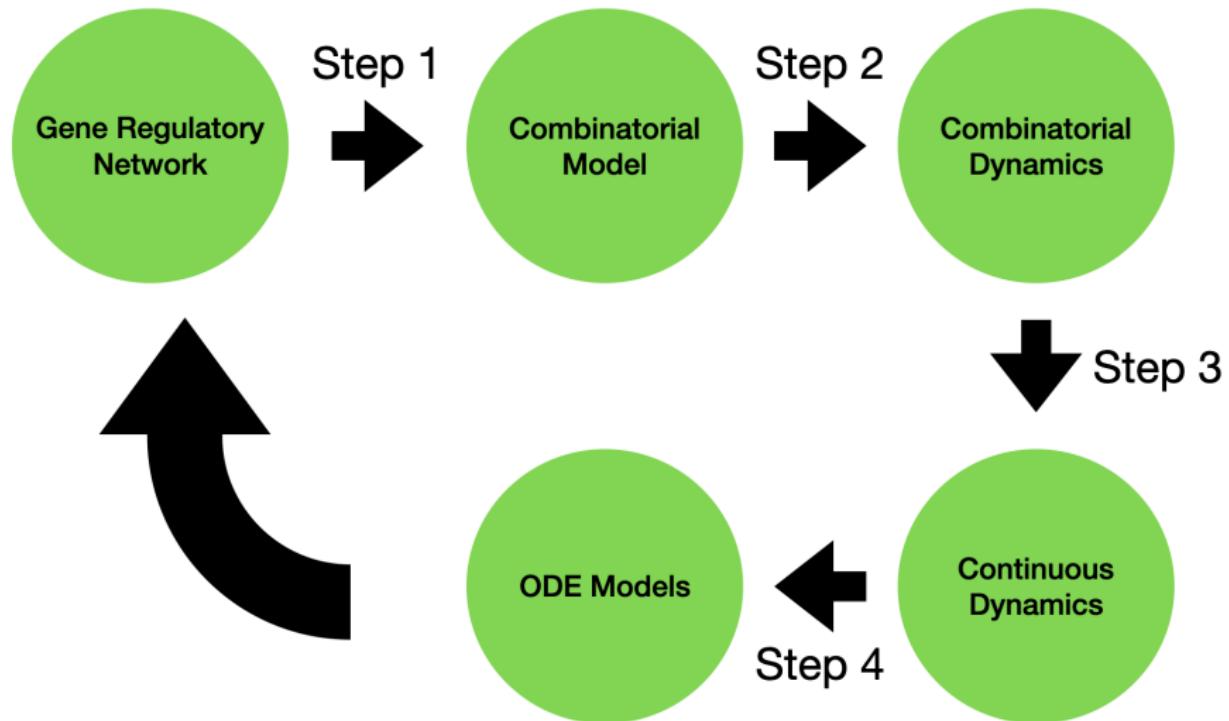


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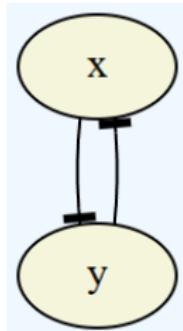
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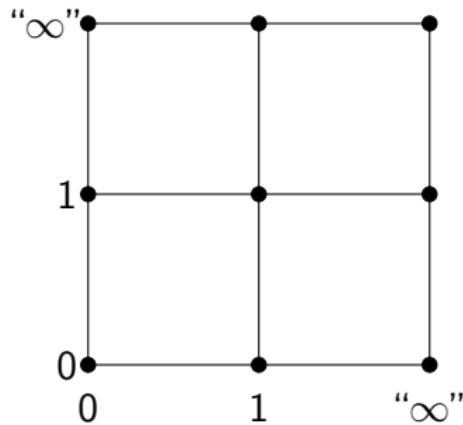
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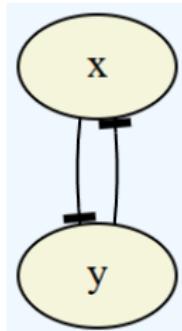
Cubical Cell Complex



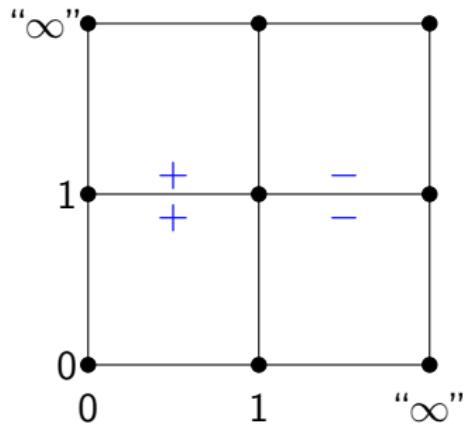
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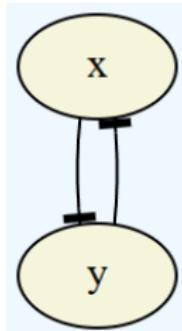
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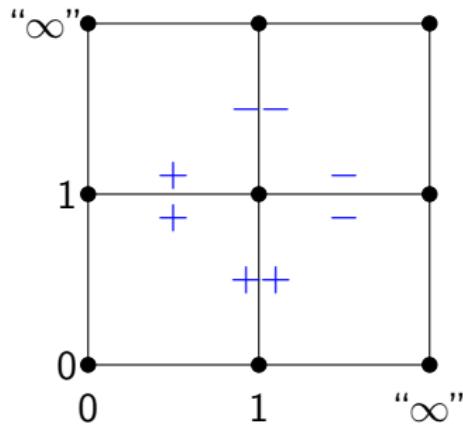
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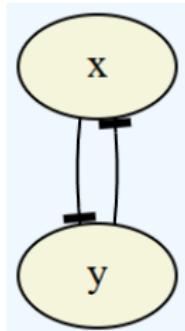
Cubical Cell Complex



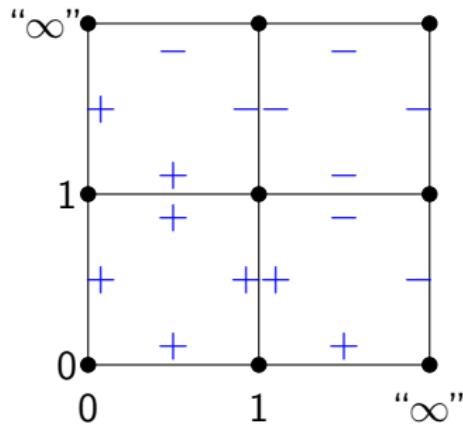
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Cubical Cell Complex



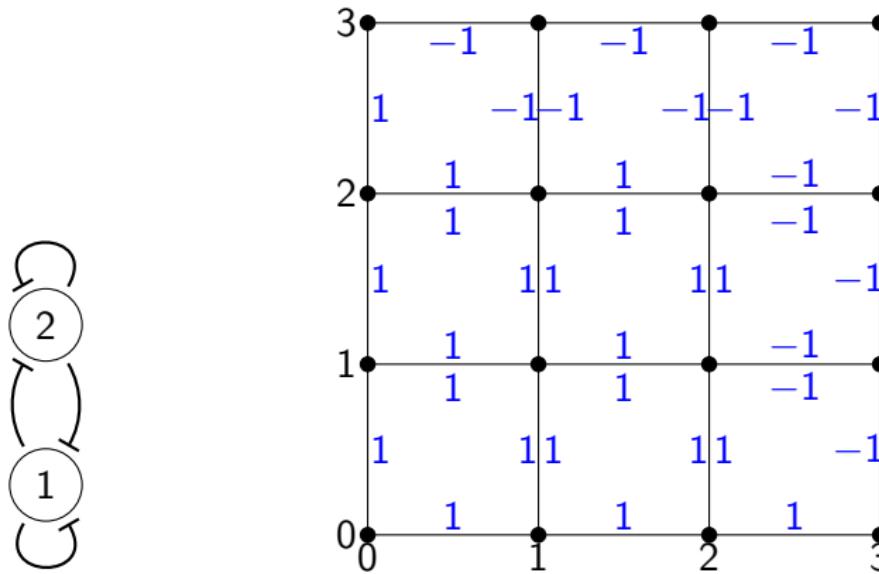
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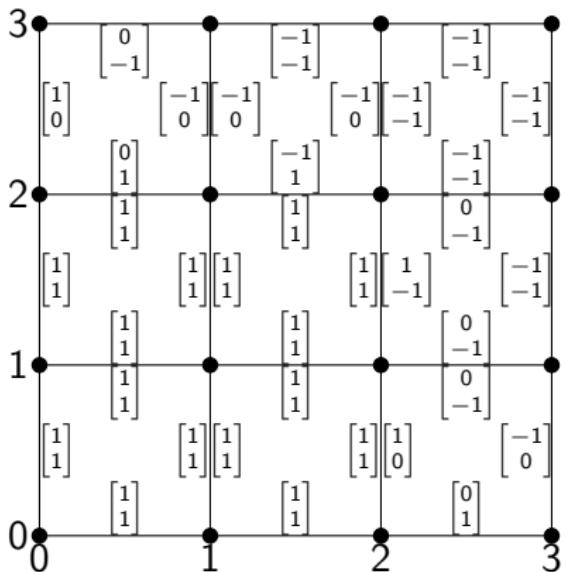
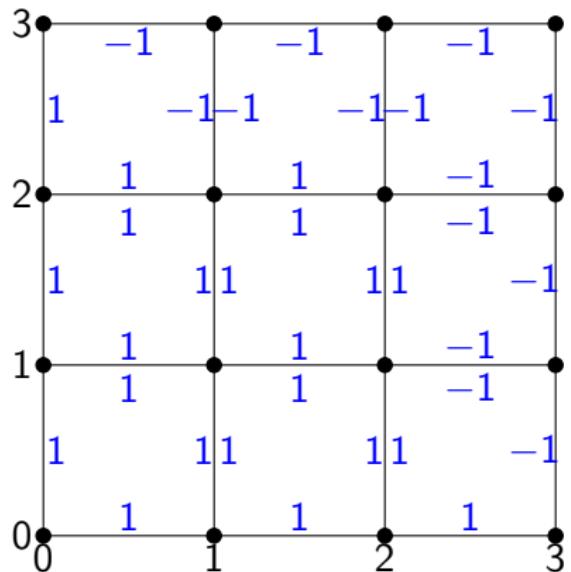
Definition

A *wall labeling* on a N -dimensional cubical complex \mathcal{X} consists of

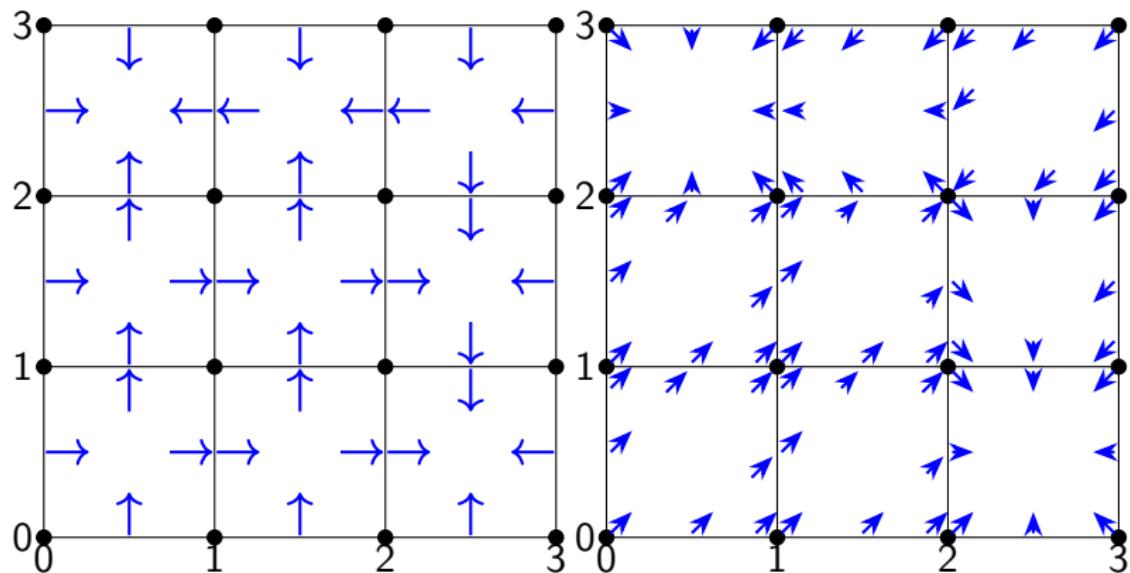
- a function $\omega: \text{TP}(\mathcal{X}) \rightarrow \{-1, +1\}$
- for each vertex $\xi \in \mathcal{X}^{(0)}$, a map $o_\xi: \{1, \dots, N\} \rightarrow \{1, \dots, N\}$ that tracks changes of ω about vertices ξ .



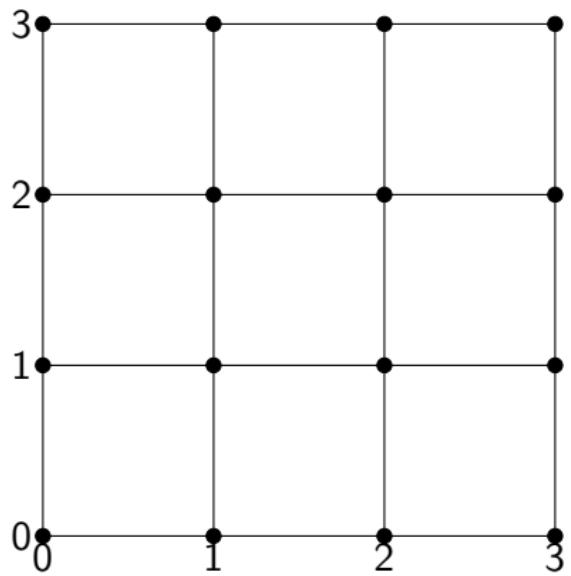
Wall labeling vs Rook Field



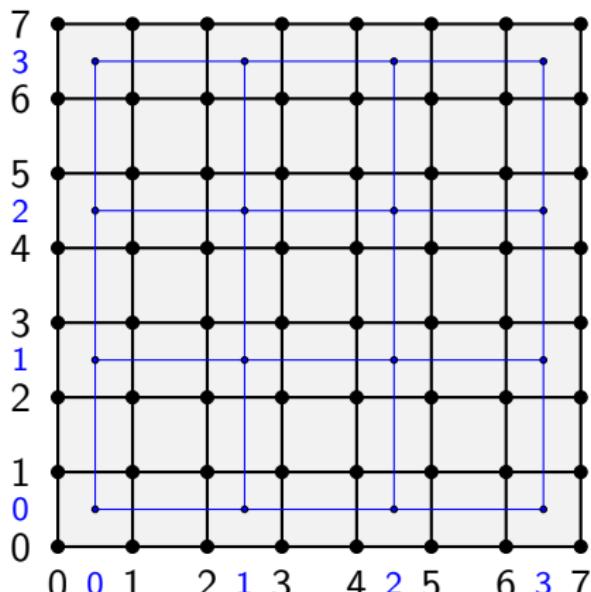
Wall labeling vs Rook Field: vector representation



Visualizing $\mathcal{F} : \mathcal{X} \rightrightarrows \mathcal{X}$ on the blow-up complex \mathcal{X}_b

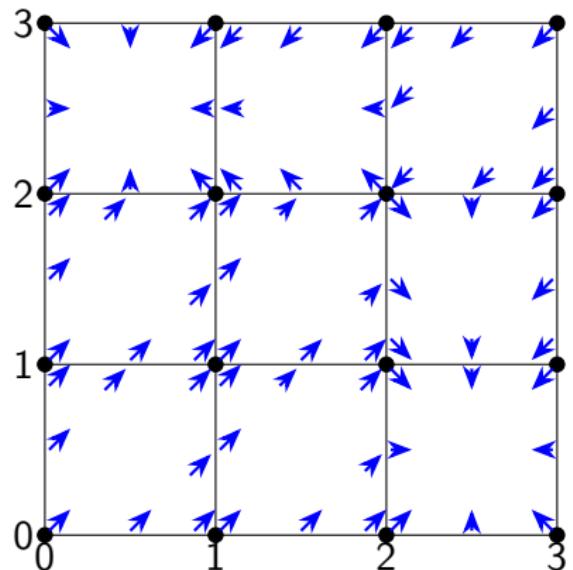


Cubical complex \mathcal{X}

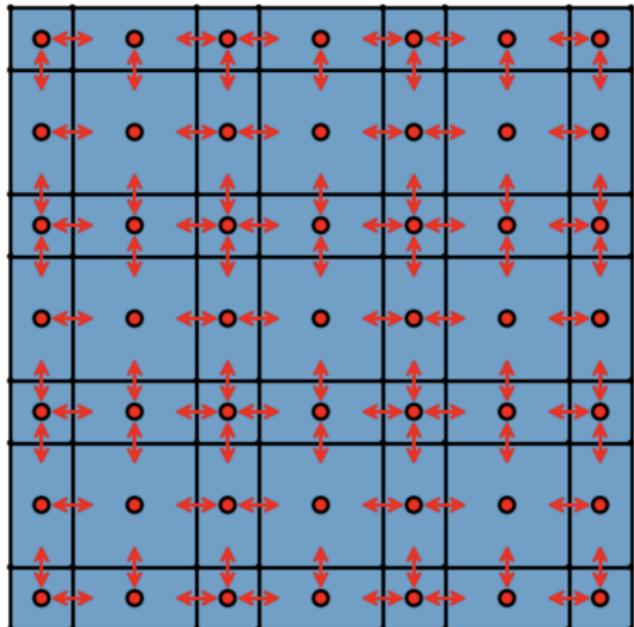


\mathcal{X} in blue, \mathcal{X}_b in black.

$$\mathcal{F}_0 : \mathcal{X} \rightrightarrows \mathcal{X}$$

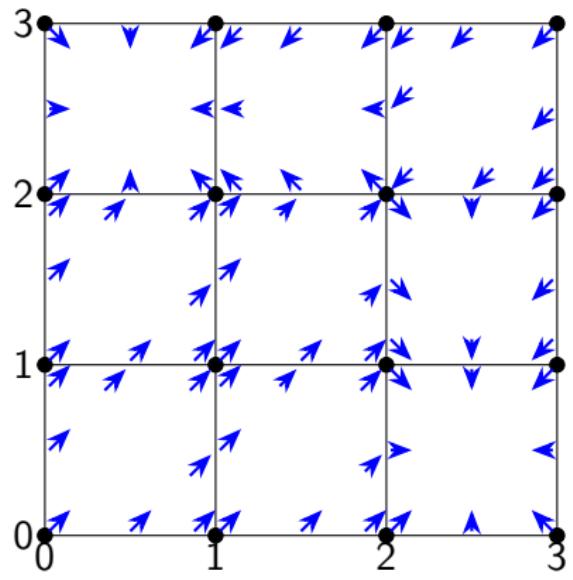


Rook Field

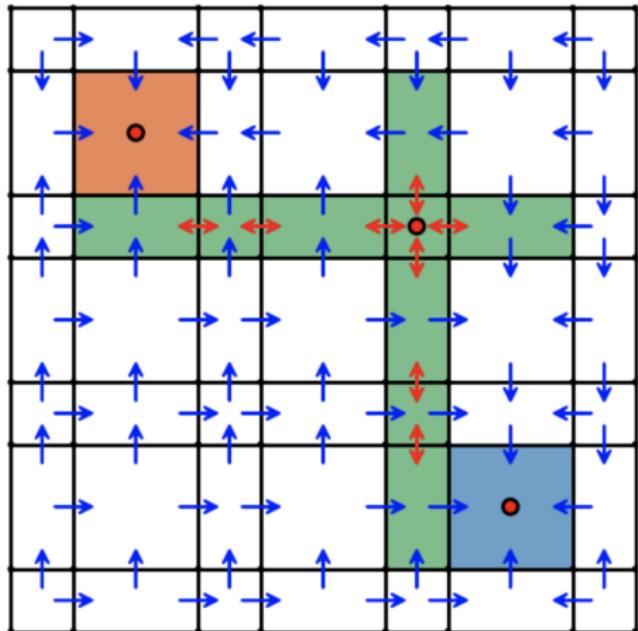


$$\mathcal{F}_0 : \mathcal{X} \rightrightarrows \mathcal{X}$$

$$\mathcal{F}_1 : \mathcal{X} \rightrightarrows \mathcal{X}$$



Rook Field

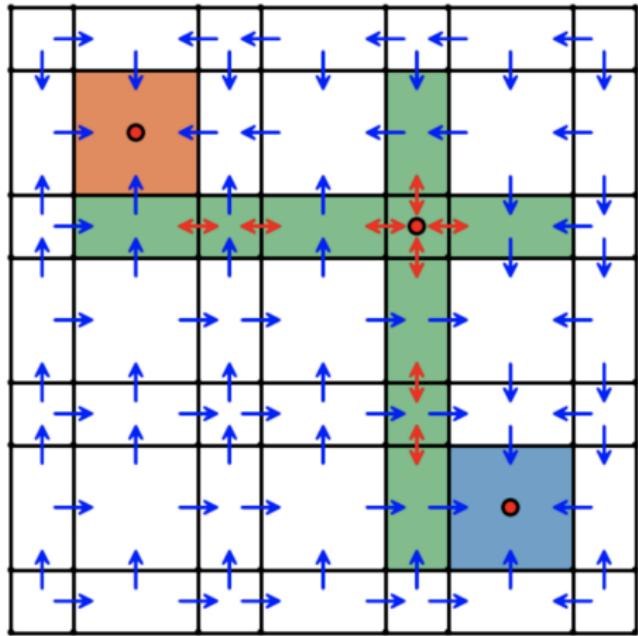


$$\mathcal{F}_1 : \mathcal{X} \rightrightarrows \mathcal{X}$$

Possible refinements of \mathcal{F}_1

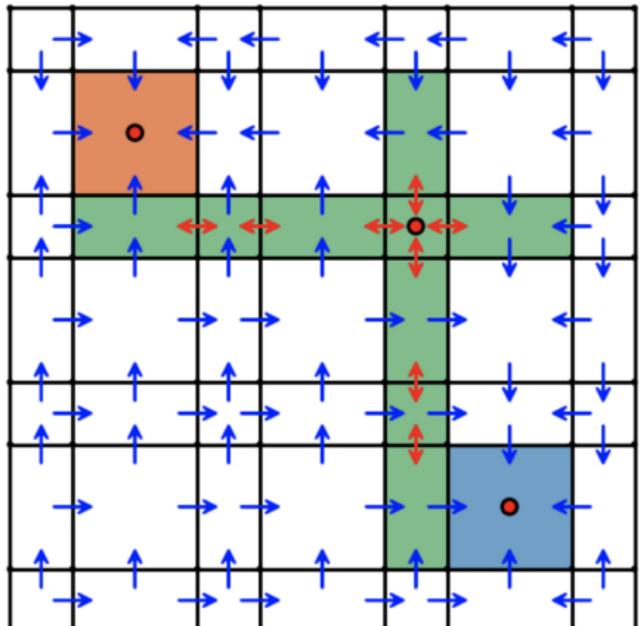
There are two types of “double-arrows” in \mathcal{F}_1 :

- non-cyclic interactions
- cyclic interactions

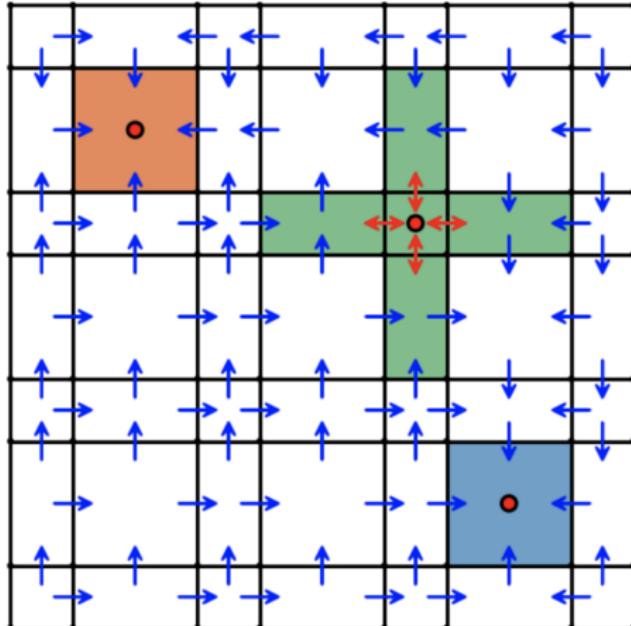


$$\mathcal{F}_1 : \mathcal{X} \rightrightarrows \mathcal{X}$$

\mathcal{F}_1 vs \mathcal{F}_2

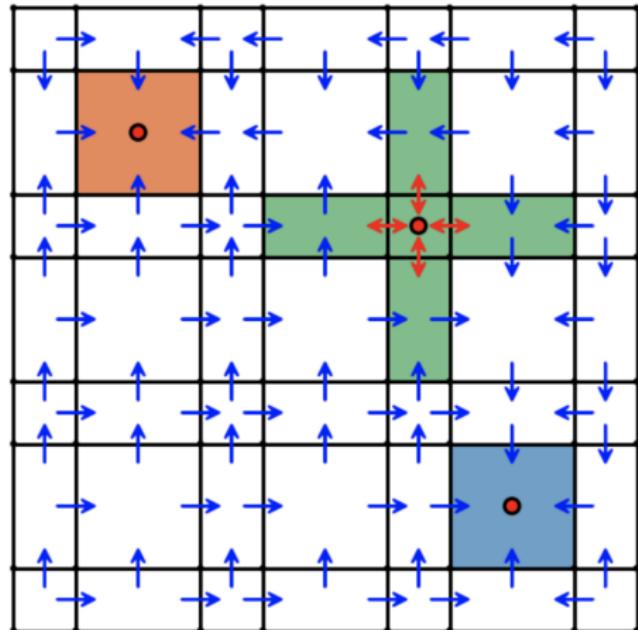
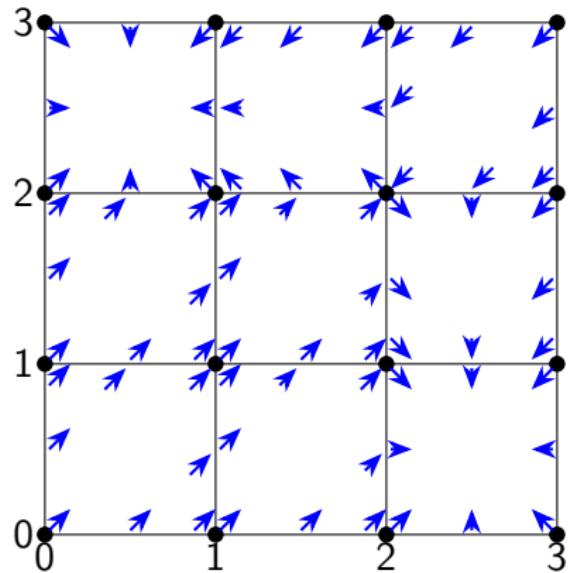


$\mathcal{F}_1 : \mathcal{X} \rightrightarrows \mathcal{X}$

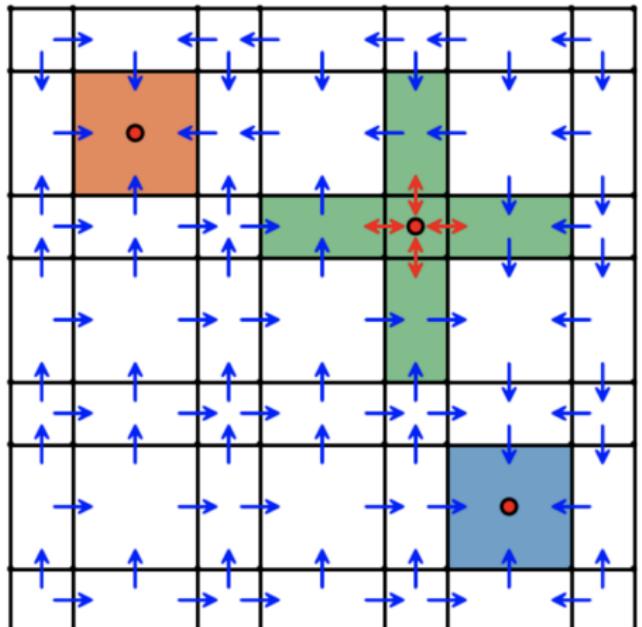


$\mathcal{F}_2 : \mathcal{X} \rightrightarrows \mathcal{X}$

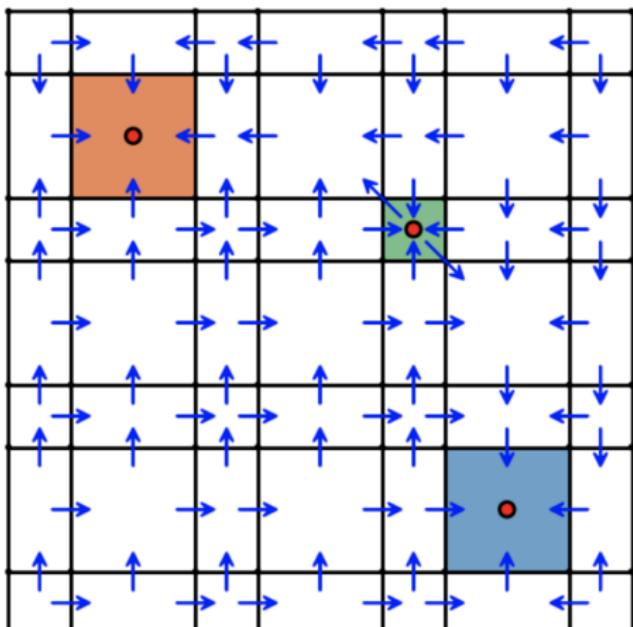
$$\mathcal{F}_2 : \mathcal{X} \rightrightarrows \mathcal{X}$$



\mathcal{F}_2 vs \mathcal{F}_3



$$\mathcal{F}_2 : \mathcal{X} \rightrightarrows \mathcal{X}$$



$$\mathcal{F}_3 : \mathcal{X} \rightrightarrows \mathcal{X}$$

Summary of Step 1

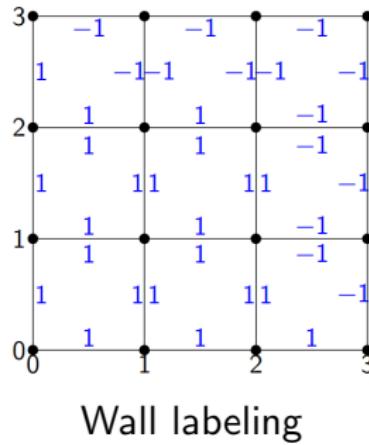


Regulatory Network

Summary of Step 1



Regulatory Network



Summary of Step 1



Regulatory Network

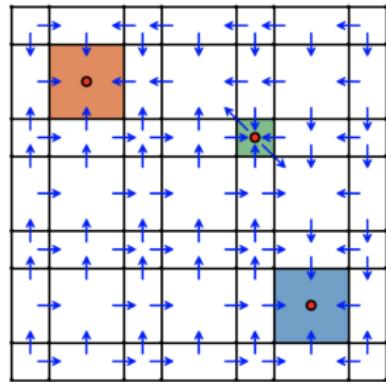
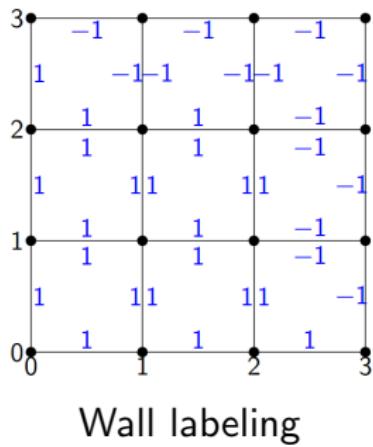


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3 Applications

Forward invariant sets

Definition

Let $\mathcal{F}: \mathcal{X} \rightrightarrows \mathcal{X}$ be a multivalued map. A set $\mathcal{N} \subseteq \mathcal{X}$ is *forward invariant* under \mathcal{F} if $\mathcal{F}(\mathcal{N}) \subseteq \mathcal{N}$.

Forward invariant sets

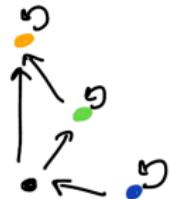
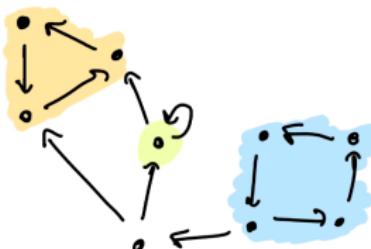
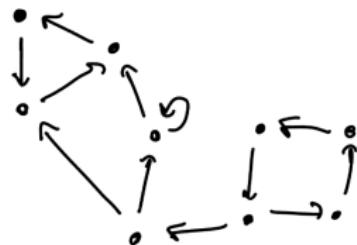
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Theorem

The collection of forward invariant sets $\text{Invset}^+(\mathcal{F})$ is a distributive lattice under the operations of intersection \cap and union \cup .

Weak condensation graph



Definition

If P is a poset, the downsets of $p \in P$ is given by $O(p) = \{q \in P \mid q \leq p\}$.

Proposition

$$O(SCC(\mathcal{F}), \leq_{\bar{\mathcal{F}}}) \cong \text{Invset}^+(\mathcal{F})$$

Retrieving P

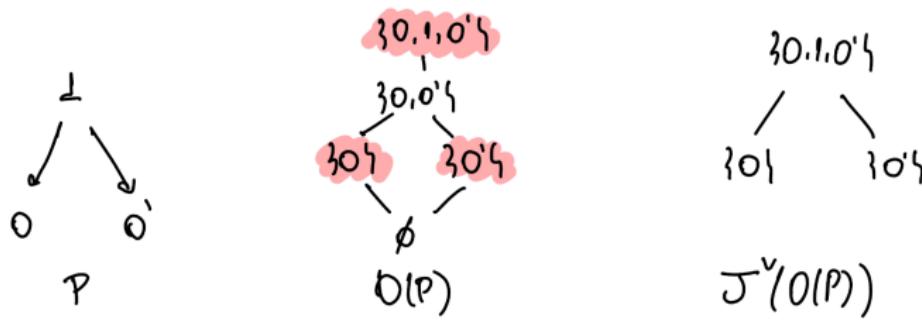
Definition (join-irreducible)

Let L be a finite distributive lattice. An element $a \in L$ is *join-irreducible* if it has a unique predecessor. The set of all join-irreducible is $J^\vee(L)$.

Theorem (Birkhoff's representation theorem)

$L \cong O(J^\vee(L))$ as lattices and $P \cong J^\vee(O(P))$ as posets.

We refer to the isomorphism as λ .



Conley Complex: combinatorial setting

Given a P -grading, that is, $\pi : \mathcal{X}_b^{(N)} \rightarrow P$, on the cubical complex \mathcal{X}_b , we can compute a P -graded chain complex on

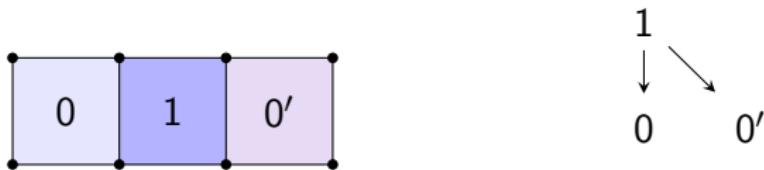
$$CH_*(p) \cong H_*(\lambda^{-1}(\mathcal{O}(p)), \lambda^{-1}(\mathcal{O}(p)^<); \mathbb{F}),$$

with boundary operator, called *connection matrix*,

$$\Delta : CH_*(p) \rightarrow \bigoplus_{q \prec p} CH_{*-1}(p).$$

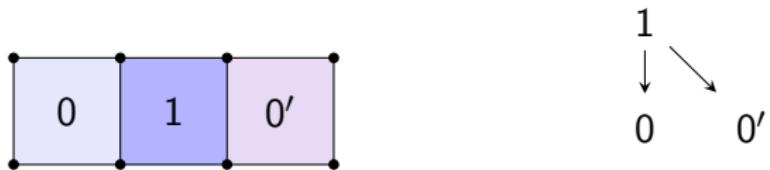
Summary of Step 2

Example:



Summary of Step 2

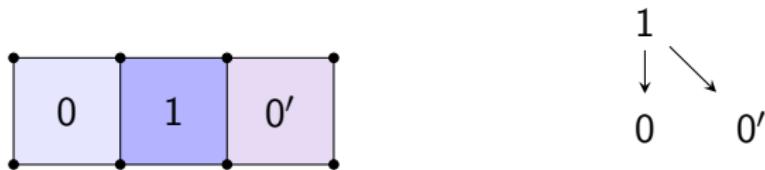
Example:



$$CH_k(0; \mathbb{F}) \cong CH_k(0'; \mathbb{F}) \cong \begin{cases} \mathbb{F} & \text{if } k = 0, \\ 0 & \text{otherwise.} \end{cases}$$

Summary of Step 2

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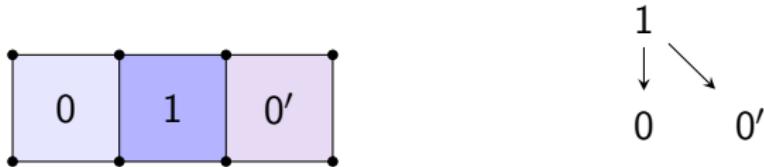


$$CH_k(0; \mathbb{F}) \cong CH_k(0'; \mathbb{F}) \cong \begin{cases} \mathbb{F} & \text{if } k = 0, \\ 0 & \text{otherwise.} \end{cases}$$

$$CH_k(1; \mathbb{F}) \cong H_k(\mathcal{X}_b, \lambda^{-1}(0) \cup \lambda^{-1}(0'); \mathbb{F}) \cong \begin{cases} \mathbb{F} & \text{if } k = 1, \\ 0 & \text{otherwise.} \end{cases}$$

Summary of Step 2

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$$\Delta_1: CH_1(1; \mathbb{F}) \rightarrow CH_0(0; \mathbb{F}) \oplus CH_0(0'; \mathbb{F}),$$

takes the form $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$

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Continuous dynamics

Let $\varphi: \mathbb{R} \times X \rightarrow X$ be a flow on a compact metric space X .

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A set $A \subset X$ is an *attractor* if there exists a compact set $K \subset X$ such that

$$A = \omega(K, \varphi) := \bigcap_{t \geq 0} \text{cl}(\varphi((t, \infty), K)) \subset \text{int}(K).$$

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Theorem (Kalies, Mischaikow, Vandervorst)

The set of all attractors of φ , $\text{Att}(\varphi)$, and the set of attracting blocks, $\text{ABlock}(\varphi)$ are bounded distributive lattices, and

$$\begin{aligned}\omega: \text{ABlock}(\varphi) &\rightarrow \text{Att}(\varphi) \\ K &\mapsto \omega(K)\end{aligned}$$

is a bounded lattice epimorphism.

Conley Complex: continuous setting

When L is a sublattice of A block that contains \emptyset and X , $J^\vee(L)$ is a Morse decomposition for φ .

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Conley Complex: continuous setting

When L is a sublattice of ABlock that contains \emptyset and X , $J^\vee(L)$ is a Morse decomposition for φ .

Let $K_0, K_1 \in \text{ABlock}(\varphi)$ and assume that $K_0 \subset K_1$.

The (*homological*) Conley index of $S = \text{Inv}(K_1 \setminus K_0, \varphi)$ is

$$CH_*(S) := H_*(K_1, K_0; \mathbb{F}).$$

Theorem (Conley)

Let S be an isolated invariant set. If $CH_*(S) \neq 0$, then $S \neq \emptyset$.

From combinatorial dynamics to continuous dynamics

Theorem

If a geometric realization $|\cdot| : \mathcal{X}_b \rightarrow \mathbb{R}^N$ satisfies certain transversality conditions with respect to the flow, then $K = |\lambda^{-1}(p)|$ is an attracting block for each $p \in P$ and the Conley complex computed in Step 2 is the Conley complex of Step 3.

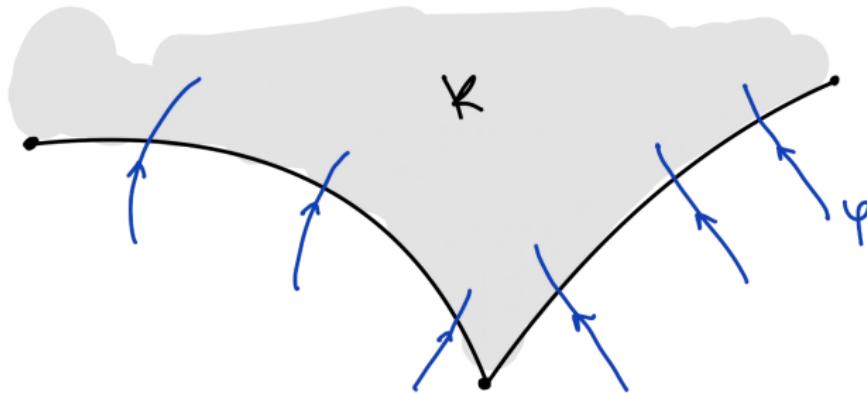


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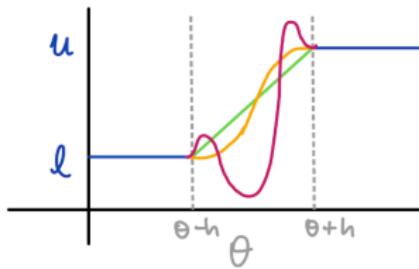
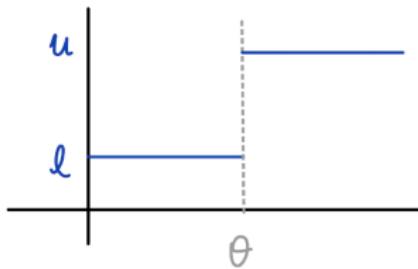
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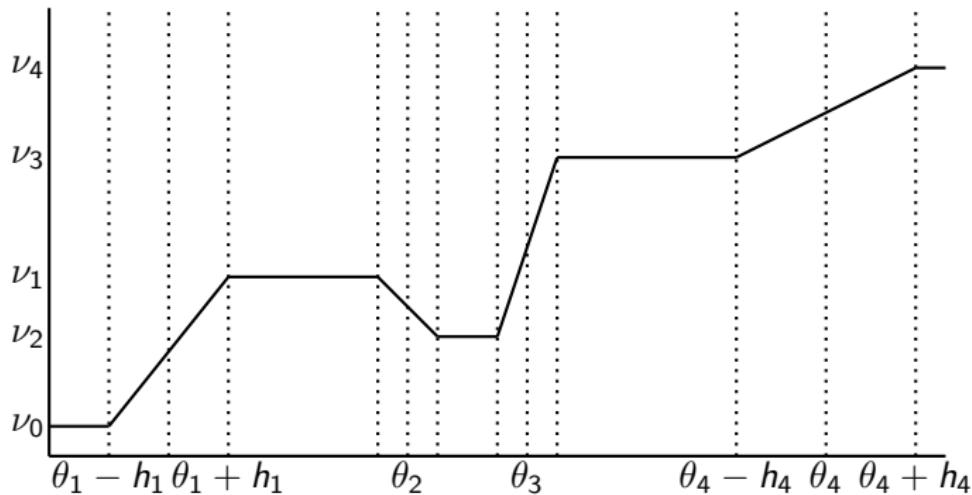
ODE models for gene expression



$$\dot{x}_n = -\gamma_n x_n + \dots$$

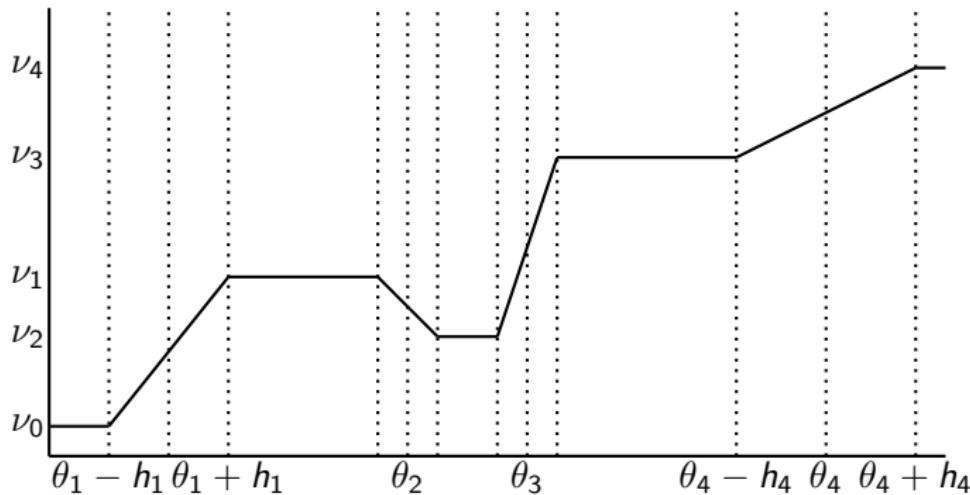
Ramp Systems

Let $\dot{x} = -\Gamma x + E(x; \nu, \theta, h)$ be a ramp system with fixed parameters.



Ramp Systems

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Remark: To go from Ramp Systems to Wall Labelings, it is enough to evaluate the vector field at $\theta \pm h$.

Validating \mathcal{F}_1

If h satisfies

$$\theta_{m_k, n, j_k} + h_{m_k, n, j_k} < \theta_{m_{k+1}, n, j_{k+1}} - h_{m_{k+1}, n, j_{k+1}}, \quad (1)$$

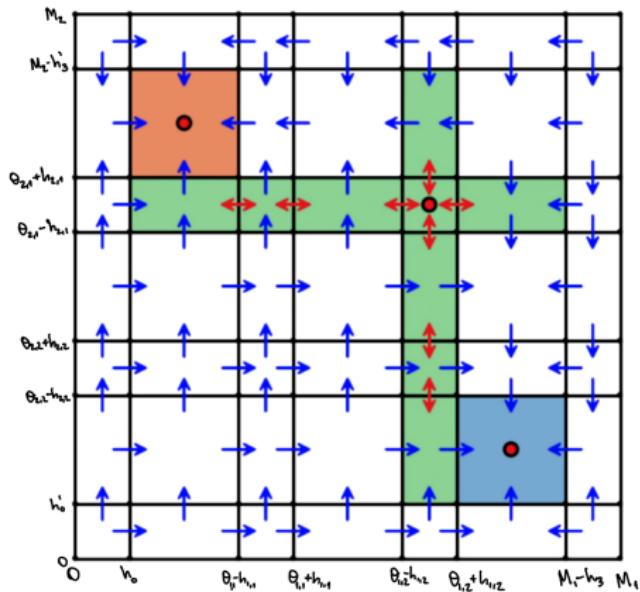
$$\frac{E_n(D_v)}{\gamma_n} \notin [\theta_{m_{k_n}, n, j_{k_n}}, \theta_{m_{k_n}, n, j_{k_n}} + h_{m_{k_n}, n, j_{k_n}}), \quad (2)$$

and

$$\frac{E_n(D_v)}{\gamma_n} \notin (\theta_{m_{k_n+1}, n, j_{k_n+1}} - h_{m_{k_n+1}, n, j_{k_n+1}}, \theta_{m_{k_n+1}, n, j_{k_n+1}}], \quad (3)$$

then there is a geometrization where the flow is transverse.

Geometric realization for \mathcal{F}_1



Validating \mathcal{F}_2 : sneak peek

It involves the construction of a perturbed surface and the following type of analytical bounds

(i) Outer transversality:

$$2h_{n_o, n_g, j_{k_{n_g}}} < \frac{L_{n_g}}{U_{n_o}} \frac{\text{length}(\xi')}{2}. \quad (4)$$

(ii) Inner transversality:

$$2h_{n_o, n_g, j_{k_{n_g}}} < \frac{L_{n_g}}{U_{n_o}} 2h_{o\xi(n_o), n_o, j_{k_{n_o}}}. \quad (5)$$

where L_n, U_n are lower/upper values that depend on the parameters.

Validating \mathcal{F}_3

Theorem

If $\dot{x} = -\Gamma x + E(x; \nu, \theta, h)$, then, in some rectangular region,

- $\dot{x}_n = -\gamma_n x_n + f_n(x_{\sigma^{-1}(n)})$ is a monotone cyclic feedback system.
- There is an unique equilibrium whose eigenvalues of the linearization satisfy

$$\chi(\lambda) = \prod(-\gamma_n - \lambda) - \frac{(-1)^k \delta C(\nu)}{\prod h_{\sigma(n), n}}$$

Validating \mathcal{F}_3

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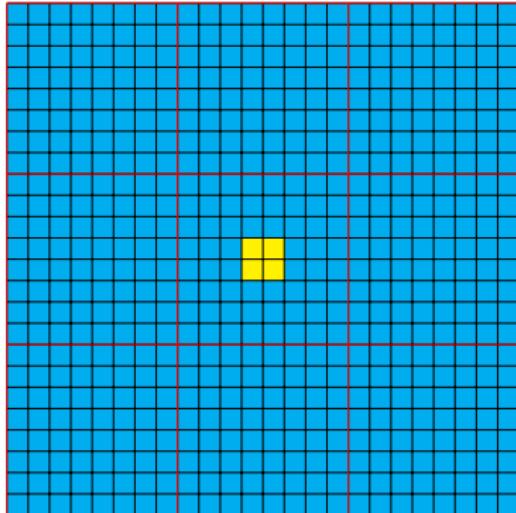
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- There is an unique equilibrium whose eigenvalues of the linearization satisfy

$$\chi(\lambda) = \prod(-\gamma_n - \lambda) - \frac{(-1)^k \delta C(\nu)}{\prod h_{\sigma(n), n}}$$

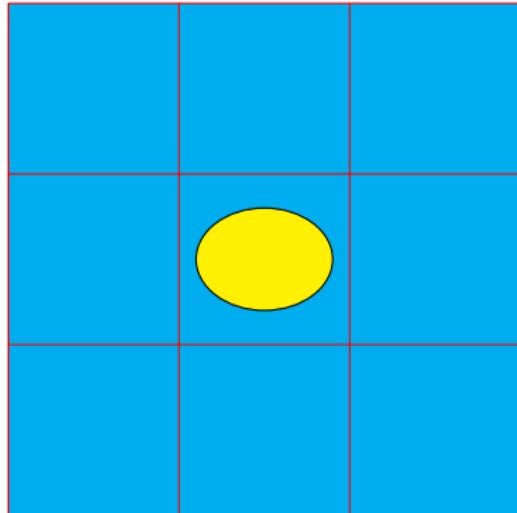
Transversality is obtained in two flavors, either by

- ① level-sets of Lyapunov functions,
- ② or similar arguments to \mathcal{F}_2 (outer/inner transversality).

Geometric realization for $N = 2$

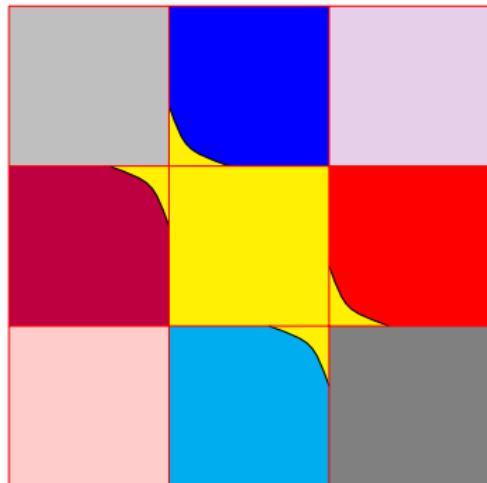
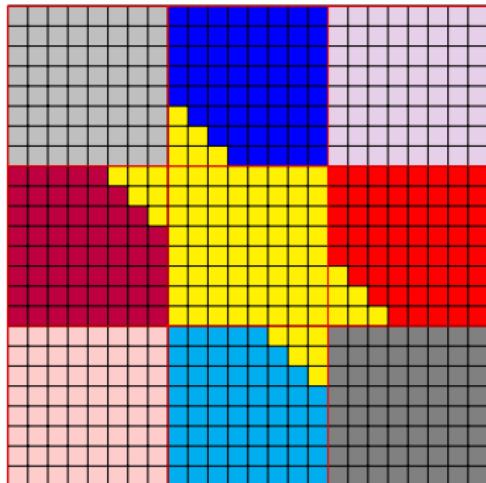


(a) \mathcal{X}_J with π_3 -grading for $\delta_{o\xi} = -1$.



(b) Geometrization of \mathcal{X}_J for $\delta_{o\xi} = -1$.

Geometric realization for $N = 2$



(b) Geometrization of \mathcal{X}_J for $\delta_{o\xi} = 1$.

The software Dynamic Signature of Gene Regulatory Networks provides **all** possible wall-labelings associated to a GRN.¹

¹up to a certain number of in and out edges.

The software Dynamic Signature of Gene Regulatory Networks provides **all** possible wall-labelings associated to a GRN.¹

This is possible by discretizing the whole parameter space into a finite collection of semi-algebraic sets.

¹up to a certain number of in and out edges.

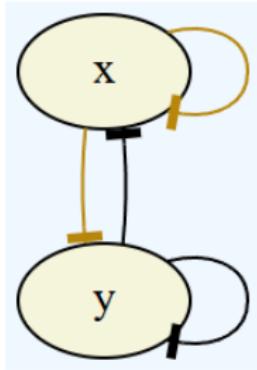
The software Dynamic Signature of Gene Regulatory Networks provides **all** possible wall-labelings associated to a GRN.¹

This is possible by discretizing the whole parameter space into a finite collection of semi-algebraic sets.

Given any region, our computation provides a complete description of its dynamics in terms of Conley Index Theory for sufficiently small h .

¹up to a certain number of in and out edges.

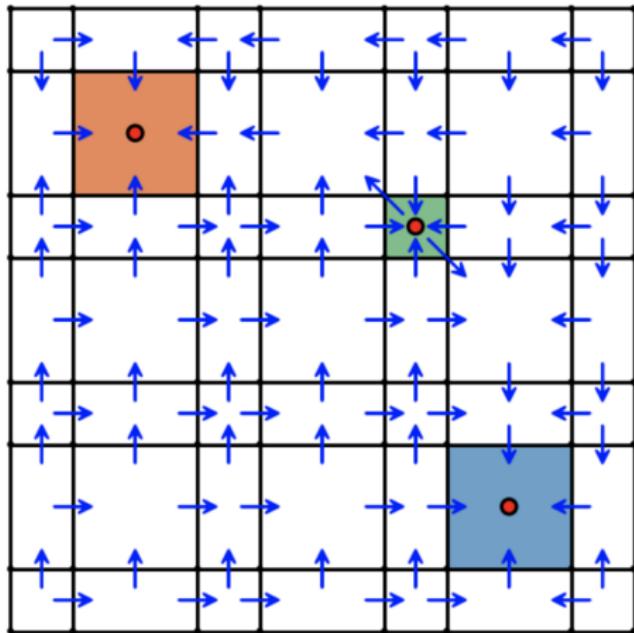
2D dimensional example: 1,600 parameter regions



Regulatory
Network

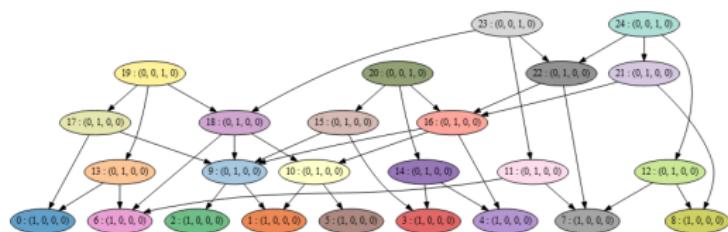
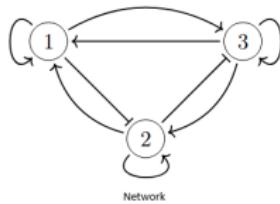
There are 1500 with dynamics given by
 $0 : (1, 0, 0)$

There are 100 with dynamics given by
 $2 : (0, 1, 0)$
 $1 : (1, 0, 0)$



3D dimensional example

$\dot{x}_1 = -\gamma_1 x_1 + r_{1,1}(x_1) + r_{1,2}(x_2) + r_{1,3}(x_3)$	$\nu_{1,1,1} = 1.01$ $\nu_{2,1,1} = 0.875$	$\nu_{1,1,2} = 4.0$ $\nu_{2,1,2} = 0.797$	$\nu_{1,2,1} = 1.0$ $\nu_{2,2,1} = 0.22$	$\nu_{1,2,2} = 4.0$ $\nu_{2,2,2} = 0.875$	$\nu_{1,3,1} = 1.0$ $\nu_{2,3,1} = 0.44$	$\nu_{1,3,2} = 2.0$ $\nu_{2,3,2} = 0.875$
$\dot{x}_2 = -\gamma_2 x_2 + r_{2,1}(x_1) (r_{2,2}(x_2) + r_{2,3}(x_3))$	$\nu_{3,1,1} = 0.76$ $\theta_{1,1} = 6.5$	$\nu_{3,1,2} = 1.0$ $\theta_{1,2} = 1.497$	$\nu_{3,2,1} = 1.0$ $\theta_{1,3} = 1.87$	$\nu_{3,2,2} = 0.85$ $\theta_{2,1} = 8.0$	$\nu_{3,3,1} = 0.5$ $\theta_{2,2} = 1.0$	$\nu_{3,3,2} = 1.0$ $\theta_{2,3} = 1.16$
$\dot{x}_3 = -\gamma_3 x_3 + r_{3,2}(x_2) (r_{3,1}(x_1) + r_{3,3}(x_3))$, $\theta_{3,1} = 3.5$	$\theta_{3,2} = 1.46$	$\theta_{3,3} = 1.61$	$\gamma_1 = \gamma_2 = 1$	$\gamma_3 = 1.2$	$h_{i,j} = 0.1$	



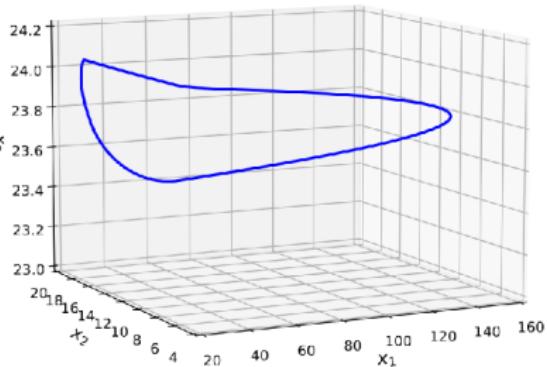
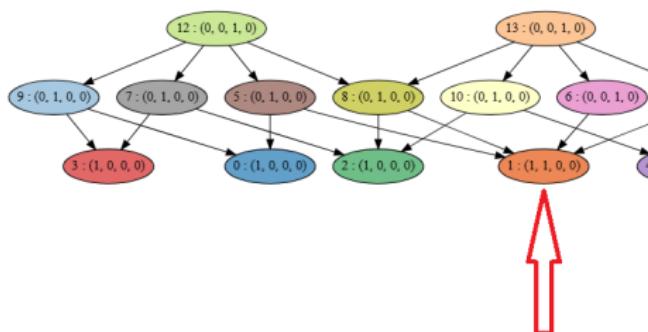
3D dimensional example

$$\dot{x}_1 = -\gamma_1 x_1 + r_{1,1}(x_1)r_{1,2}(x_2)r_{1,3}(x_3)$$

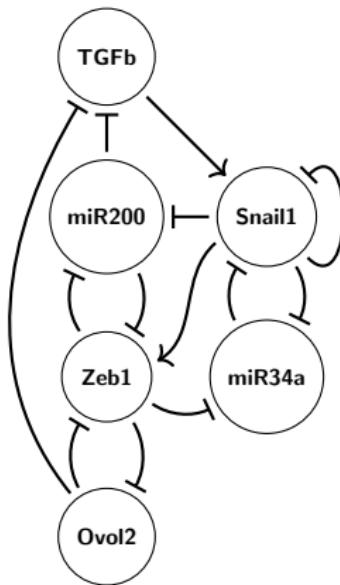
$$\dot{x}_2 = -\gamma_2 x_2 + r_{2,1}(x_1)r_{2,2}(x_2)$$

$$\dot{x}_3 = -\gamma_3 x_3 + r_{3,2}(x_2)r_{3,3}(x_3),$$

$\nu_{1,1,1} = 1.80$	$\nu_{1,1,2} = 8.56$	$\nu_{1,2,1} = 13.07$	$\nu_{1,2,2} = 3.25$	$\nu_{1,3,1} = 20.10$	$\nu_{1,3,2} = 1.07$
$\nu_{2,1,1} = 2.44$	$\nu_{2,1,2} = 0.84$	$\nu_{2,2,1} = 0.16$	$\nu_{2,2,2} = 6.10$	$\nu_{3,2,1} = 2.39$	$\nu_{3,2,2} = 1.36$
$\nu_{3,3,1} = 0.05$	$\nu_{3,3,2} = 5.03$	$\theta_{1,1} = 27.17$	$\theta_{1,2} = 2.26$	$\theta_{1,3} = 11.73$	$\theta_{2,1} = 39.10$
$\theta_{2,2} = 1.25$	$\theta_{3,2} = 10.47$	$\theta_{3,3} = 6.70$	$\gamma_1 = 1$	$\gamma_2 = \gamma_3 = 0.5$	$h_{i,j} = 0.5$



6D dimensional example



Regulatory network for reversible epithelial-to-mesenchymal transition as proposed by T. Hong, K. Watanabe, C.H. Ta, A. Villarreal-Ponce, Q. Nie, and X. Dai, An *ovol2-zeb1* mutual inhibitory circuit governs epithelial-mesenchymal transitions, PLOS Comput. Biol. 11 (2015)

58 dim parameter space

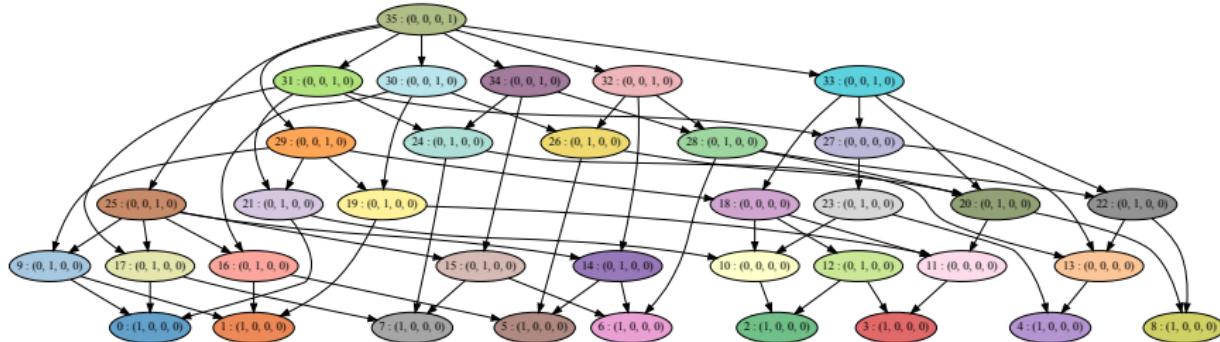


Figure: Morse graph for the map \mathcal{F}_3 generated by the parameter node 52,717,613,010.

58 dim parameter space and 4,429,771,960,320 parameter regions.

Parameter node 1, 739, 757, 491, 101

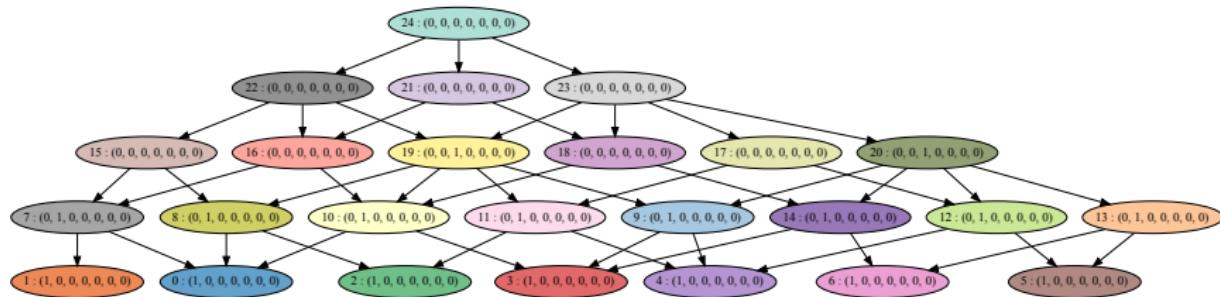


Figure: Morse graph for the map \mathcal{F}_2 generated by the parameter node 1, 739, 757, 491, 101 of the parameter graph of the network in Figure 43.

In this example there are 17 equilibria, all identified by the morse sets.

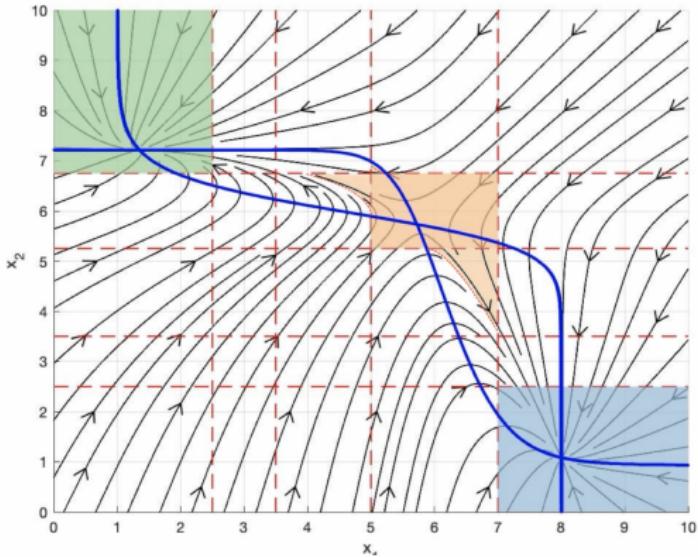
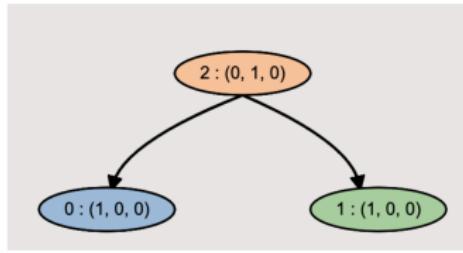
Hill Systems

Hill functions

$$\begin{aligned}\dot{x}_1 &= -\gamma_1 x_1 + H_{1,1}^-(x_1)H_{1,2}^-(x_2) \\ \dot{x}_2 &= -\gamma_2 x_2 + H_{2,1}^-(x_1)H_{2,2}^-(x_2),\end{aligned}$$

$$H_{i,j}^-(x) = \nu_{i,j,2} + (\nu_{i,j,1} - \nu_{i,j,2}) \frac{\theta_{i,j}^{s_{i,j}}}{x^{s_{i,j}} + \theta_{i,j}^{s_{i,j}}},$$

$\nu_{1,1,1} = 1.8662, \nu_{1,1,2} = 1.1061, \theta_{1,1} = 0.6927$
 $\nu_{1,2,1} = 1.1705, \nu_{1,2,2} = 0.2608, \theta_{1,2} = 1.9914$
 $\nu_{2,1,1} = 2.8064, \nu_{2,1,2} = 0.1614, \theta_{2,1} = 1.1436$
 $\nu_{2,2,1} = 0.7340, \nu_{2,2,2} = 0.6581, \theta_{2,2} = 0.4310$
 $\gamma_1 = 0.8147, \gamma_2 = 0.9058, s_{i,j} = 20.$



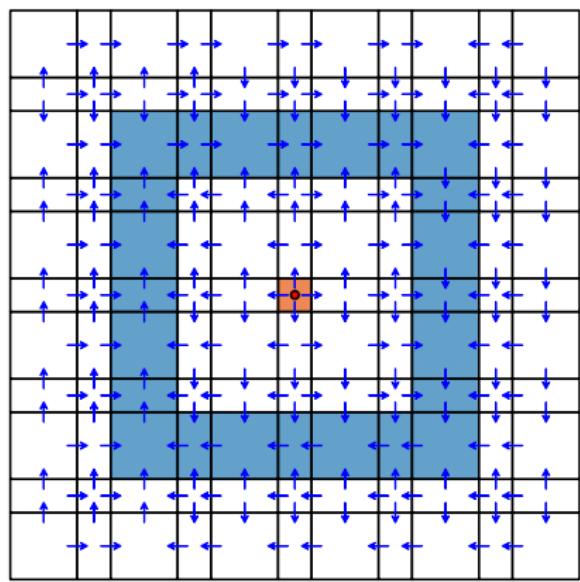
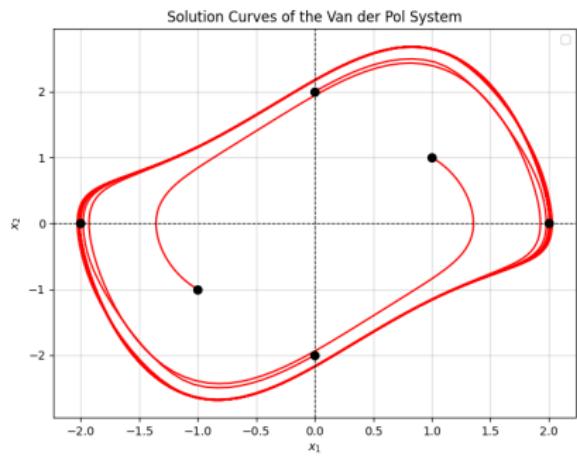
Ramp Approximations of ODEs: Van der Pol oscillator

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -x_1 + (1 - x_1^2)x_2$$

$$\dot{x}_1 = -\gamma_1 x_1 + r_{1,1}(x_1) + r_{1,2}(x_2)$$

$$\dot{x}_2 = -\gamma_2 x_2 + r_{2,1}(x_1) + r_{2,2}(x_2)$$



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Questions

Thank you for your attention!
Questions?