

# Robust chaos in piecewise-linear maps

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Indranil Ghosh

School of Mathematics and statistics, University College Dublin

David J.W. Simpson

School of Mathematical and Computational Sciences, Massey University

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University College Dublin

Email: [indra.ghosh@ucd.ie](mailto:indra.ghosh@ucd.ie)

# Importance

1. Piecewise-linear maps have applications in designing secure encryption schemes (one of the main motivations that drives me!).
2. Investigating chaos regions might let engineers design proper fail-safes in switched systems in avionics, for example!
3. Prevention of undesirable chaotic regimes while designing DC-DC power converters and inverters.



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4. Many more “..., which this margin is too small to contain.”



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# Border-collision normal form (BCNF)

1. In our project we study the 2D BCNF (Nusse & Yorke, 1992)

$$f_\xi(x, y) = \begin{cases} \begin{bmatrix} \tau_L & 1 \\ -\delta_L & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}, & x \leq 0, \\ \begin{bmatrix} \tau_R & 1 \\ -\delta_R & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}, & x \geq 0. \end{cases}$$

2. Here  $(x, y) \in \mathbb{R}^2$  and  $\xi = (\tau_L, \delta_L, \tau_R, \delta_R) \in \mathbb{R}^4$  are the parameters.



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2. Here  $(x, y) \in \mathbb{R}^2$  and  $\xi = (\tau_L, \delta_L, \tau_R, \delta_R) \in \mathbb{R}^4$  are the parameters.

3. Any continuous, two-piece, piecewise-linear map on  $\mathbb{R}^2$  satisfying a certain non-degeneracy condition can be converted to it.



source: <https://tintin.fandom.com/wiki/Tintin>

# Phase portrait of a chaotic attractor

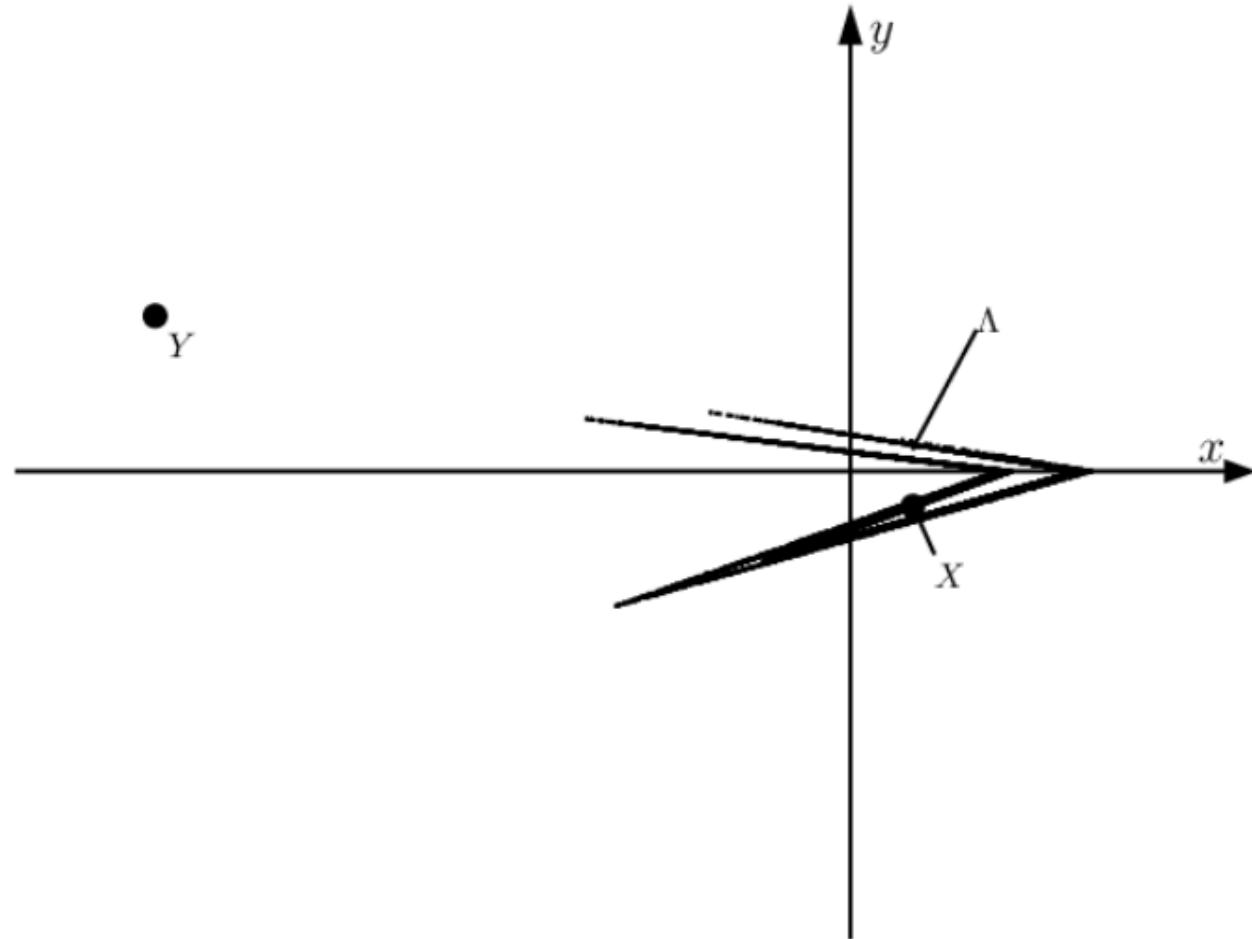


Figure: A sketch of the phase portrait of  $f_\xi$  with  $\xi \in \Phi_{\text{BYG}}$

# Renormalisation operator

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1. Renormalisation: for some members of a family of maps, a higher iterate or induced map is conjugate to different member of this family.
2. Although the second iterate  $f_\xi^2$  has four pieces, relevant dynamics only occurs in two of these:

$$f_\xi^2(x, y) = \begin{cases} \begin{bmatrix} \tau_L \tau_R - \delta_L & \tau_R \\ -\delta_R \tau_L & -\delta_R \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} \tau_R + 1 \\ -\delta_R \end{bmatrix}, & x \leq 0, \\ \begin{bmatrix} \tau_R^2 - \delta_R & \tau_R \\ -\delta_R \tau_R & -\delta_R \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} \tau_R + 1 \\ -\delta_R \end{bmatrix}, & x \geq 0. \end{cases}$$

# Renormalisation operator

3. Now  $f_\xi^2$  can be transformed to  $f_{g(\xi)}$ , where  $g$  is the **renormalisation operator** (Ghosh and Simpson, 2022)  $g : \mathbb{R}^4 \mapsto \mathbb{R}^4$ , given by

$$(\tilde{\tau}_L, \tilde{\delta}_L, \tilde{\tau}_R, \tilde{\delta}_R) = (\tau_R^2 - 2\delta_R, \delta_R^2, \tau_L\tau_R - \delta_L - \delta_R, \delta_L\delta_R).$$

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4. We perform a coordinate change to put  $f_\xi^2$  in the normal form:

$$\begin{bmatrix} \tilde{x}' \\ \tilde{y}' \end{bmatrix} = \begin{cases} \begin{bmatrix} \tilde{\tau}_L & 1 \\ -\tilde{\delta}_L & 0 \end{bmatrix} \begin{bmatrix} \tilde{x} \\ \tilde{y} \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}, & \tilde{x} \leq 0, \\ \begin{bmatrix} \tilde{\tau}_R & 1 \\ -\tilde{\delta}_R & 0 \end{bmatrix} \begin{bmatrix} \tilde{x} \\ \tilde{y} \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}, & \tilde{x} \geq 0. \end{cases}$$

# Results

1. We consider the parameter region

$$\Phi = \{ \xi \in \mathbb{R}^4 \mid \tau_L > \delta_L + 1, \delta_L > 0, \tau_R < -(\delta_R + 1), \delta_R > 0 \}.$$

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4. The attractor is often destroyed at  $\phi^+(\xi) = 0$ : homoclinic bifurcation (Banerjee, Yorke & Grebogi, 1998). Thus focused on the region

$$\Phi_{\text{BYG}} = \{ \xi \in \Phi \mid \phi^+(\xi) > 0 \}.$$

# Results

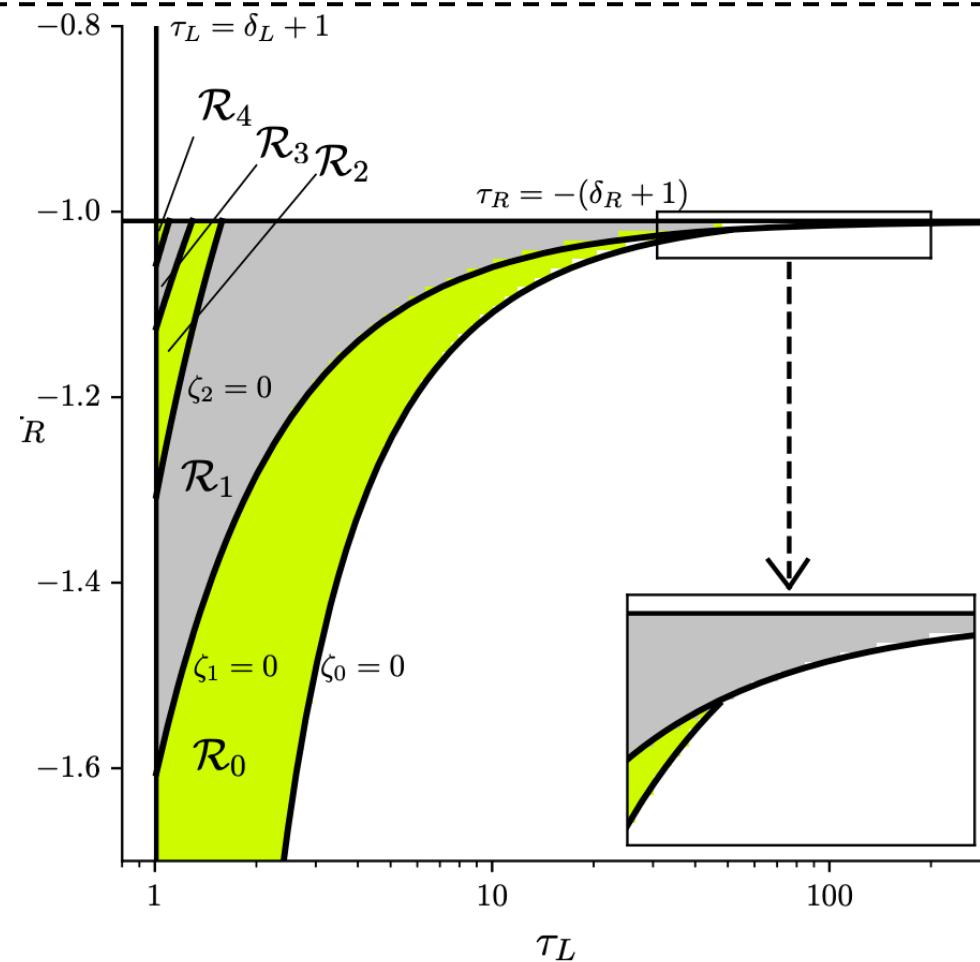


Figure: Sketch of two-dimensional cross-section of  $\Phi_{\text{BYG}}$  when  $\delta_L = \delta_R = 0.01$

# Results

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Theorem (Ghosh & Simpson, 2022)

*The  $\mathcal{R}_n$  are non-empty, mutually disjoint, and converge to the fixed point (1, 0, -1, 0) as  $n \rightarrow \infty$ . Moreover*

$$\Phi_{\text{BYG}} \subset \bigcup_{n=0}^{\infty} \mathcal{R}_n.$$

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Theorem (Ghosh & Simpson, 2022)

*With any  $\xi \in \mathcal{R}_0$ ,  $\Lambda(\xi)$  is bounded, connected, and invariant. Moreover, it is chaotic (positive Lyapunov exponent).*

# Results

Theorem (Ghosh & Simpson, 2022)

For any  $\xi \in \mathcal{R}_n$ , where  $n \geq 0$ ,  $g^n(\xi) \in \mathcal{R}_0$  and there exist mutually disjoint sets  $S_0, S_1, \dots, S_{2^n-1} \subset \mathbb{R}^2$  such that  $f_\xi(S_i) = S_{(i+1) \bmod 2^n}$  and

$f_\xi^{2^n}|S_i$  is affinely conjugate to  $f_{g^n(\xi)}|\Lambda(g^n(\xi))$

for each  $i \in \{0, 1, \dots, 2^n - 1\}$ . Moreover,

$$\bigcup_{i=0}^{2^n-1} S_i = \text{cl}\left(W^u(\gamma_n)\right),$$

where  $\gamma_n$  is a saddle-type periodic solution of our map having the symbolic itinerary  $\mathcal{F}^n(R)$ :

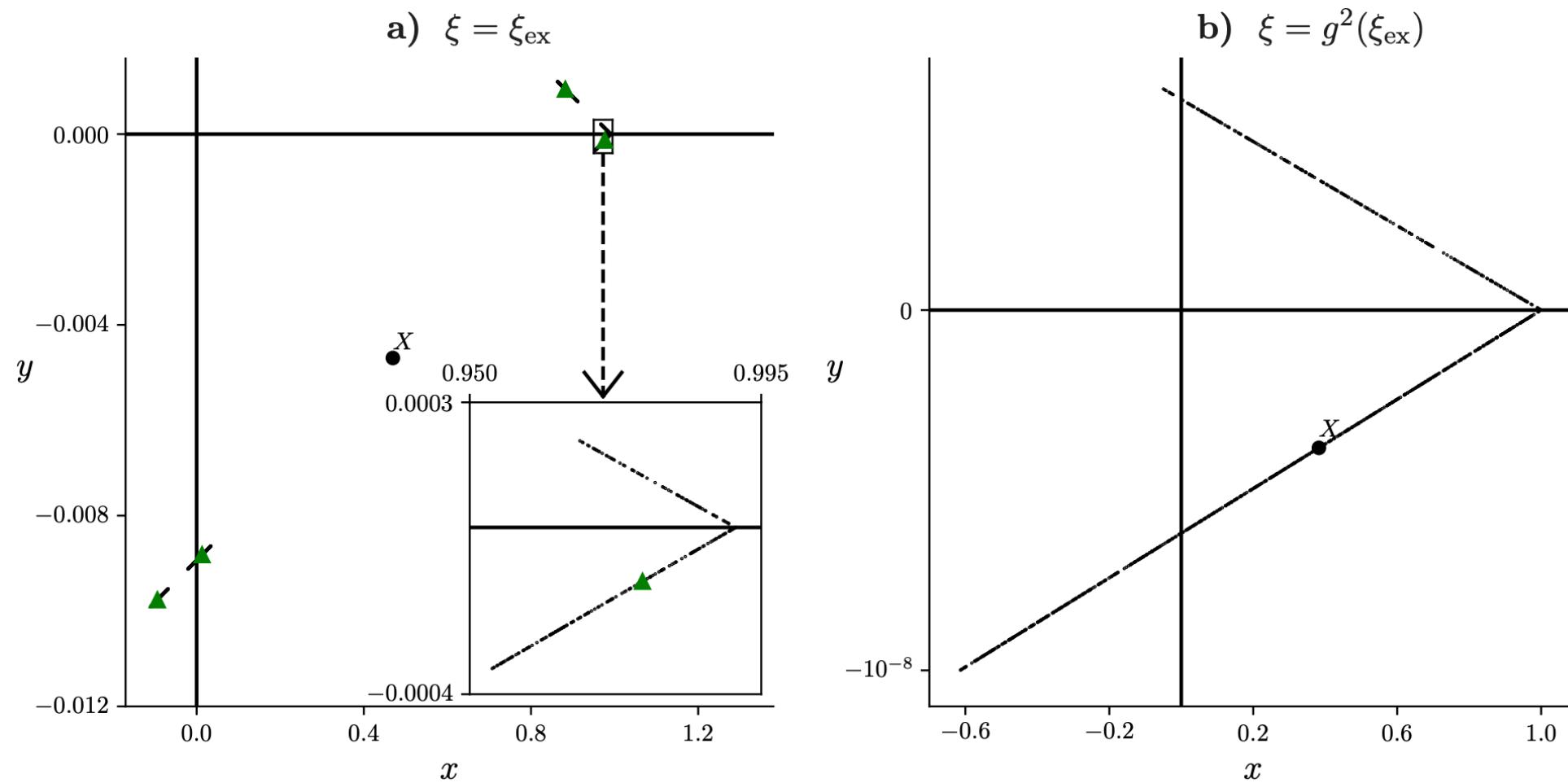
# Results

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$n$	$\mathcal{F}^n(R)$
0	R
1	LR
2	RRLR
3	LRLRRRLR
4	RRLRRLRLRLRRRLR

Table: The first 5 words in the sequence generated by the repeatedly applying the substitution rule  $(L, R) \mapsto (RR, LR)$  to  $\mathcal{W} = R$ .

# Results



# Devaney chaos

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Theorem (Ghosh & Simpson, 2022)

*Let  $\xi \in \Phi_{\text{BYG}}$  and suppose  $J_1(\xi) > 1$  and  $\lambda_L^s + |\lambda_R^s| < 1$ . Then  $W^s(X)$  is dense in a triangular region containing  $\Lambda$ .*

# Devaney chaos

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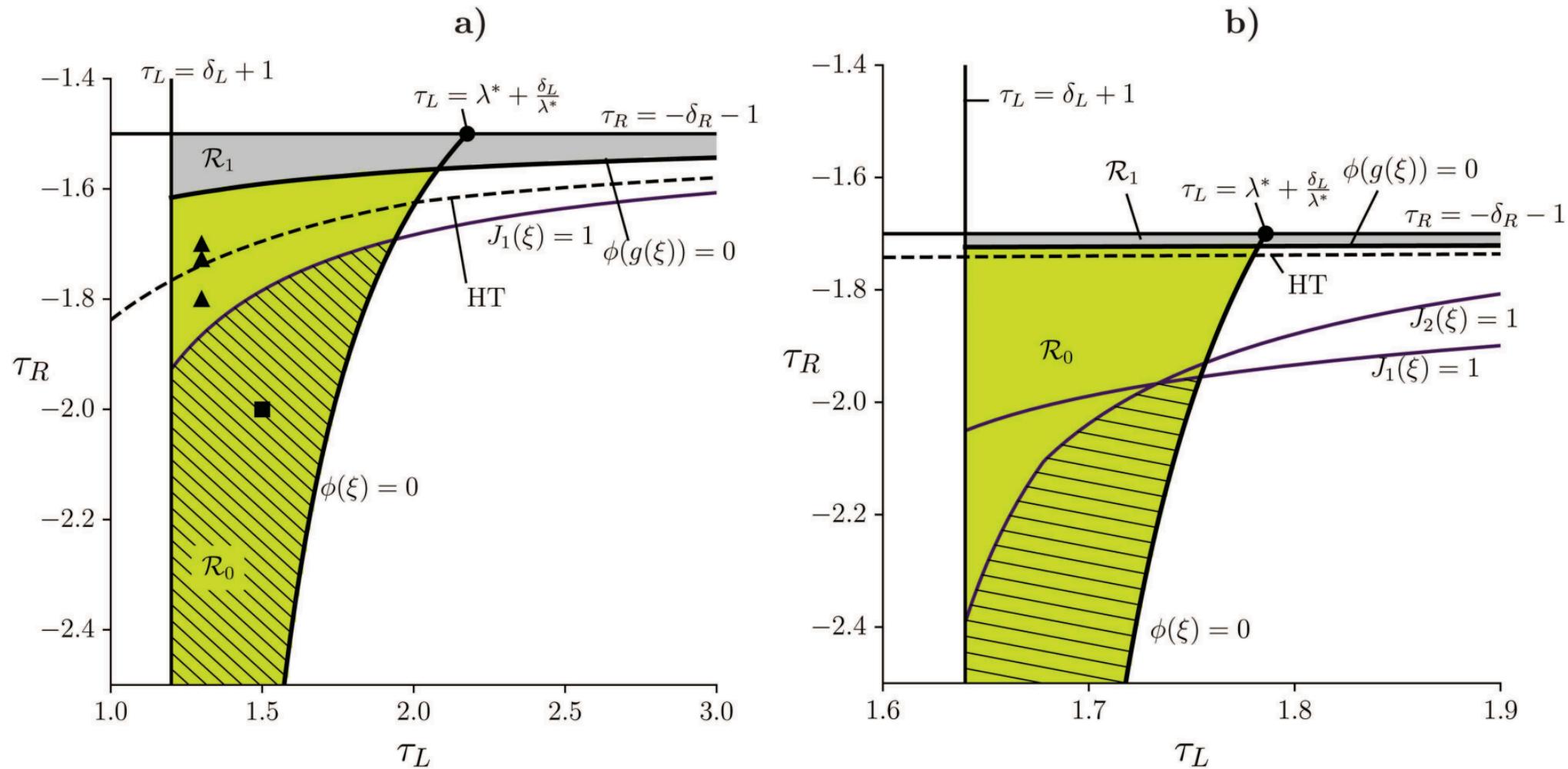
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Theorem (Ghosh & Simpson, 2022)

*Let  $\xi \in \Phi_{\text{BYG}}$  and suppose  $J_1(\xi) > 1$  and  $J_2(\xi) < 1$ . Then  $f_\xi$  is chaotic in the sense of Devaney on  $\Lambda$ .*

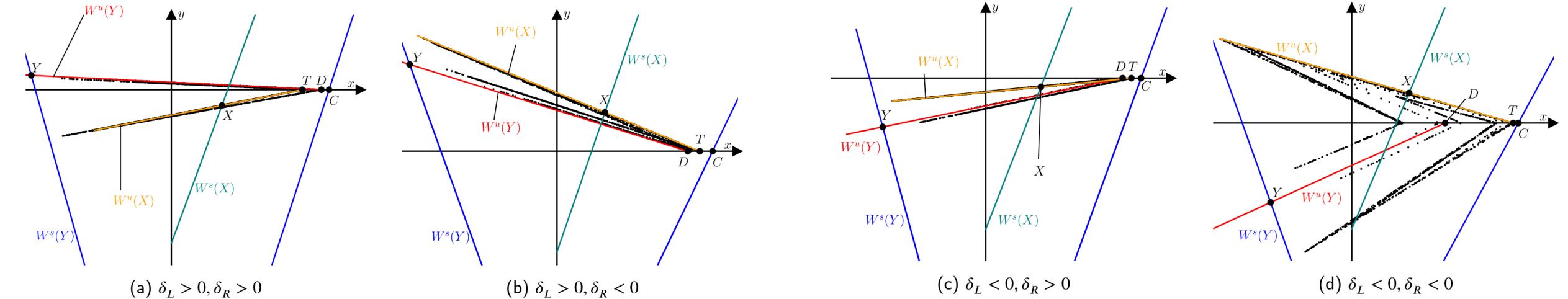
# Devaney Chaos



# Generalised parameter region

Now we consider the more generalised parameter region considering the orientation-reversing and non-invertible cases,

$$\Phi = \{ \xi \in \mathbb{R}^4 \mid \tau_L > |\delta_L + 1|, \tau_R < -|\delta_R + 1| \}.$$



# Invariant expanding cone

Chaos in  $\Phi_{\text{BYG}}$  can be proved by constructing an invariant expanding cone in tangent space (Glendinning & Simpson, 2021). We have extended this to  $\Phi$ .

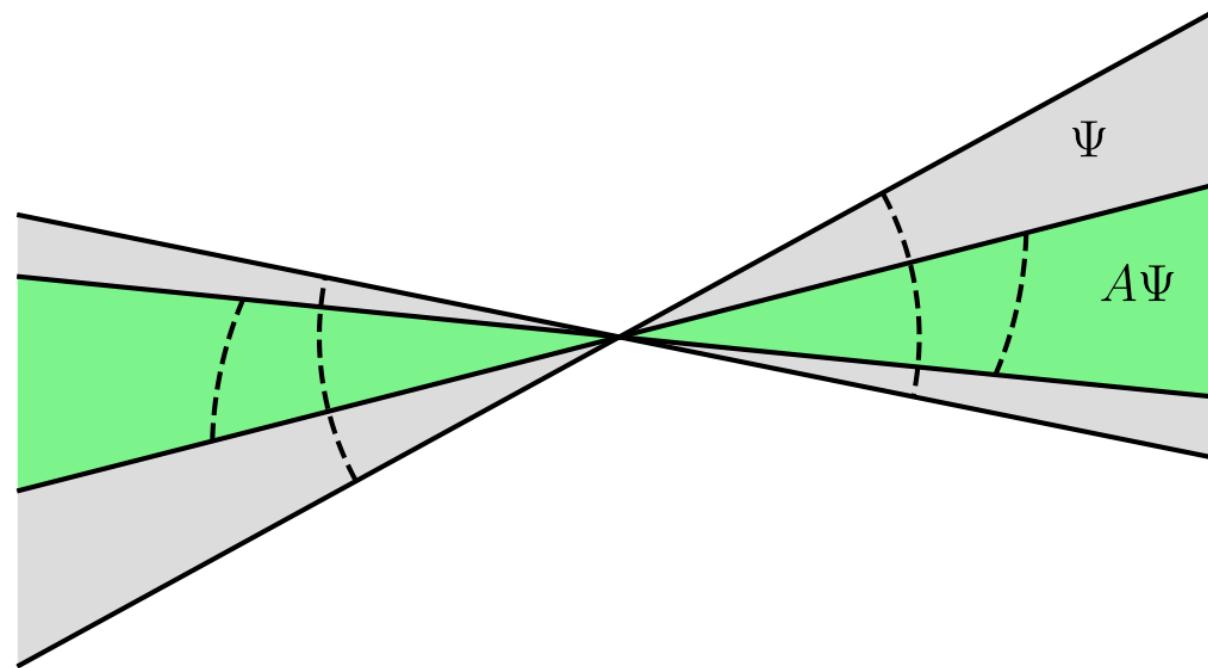


Figure: A sketch of an invariant expanding cone  $\Psi$  and its image  $A\Psi = \{Av \mid v \in \Psi\}$ , given  $A \in \mathbb{R}^{2 \times 2}$ .

# Robust chaos in a generalised setting

Theorem (Ghosh, McLachlan, & Simpson, 2023)

*For any  $\xi \in \Phi_{\text{trap}} \cap \Phi_{\text{cone}}$ , the normal form has a topological attractor with a positive Lyapunov exponent.*

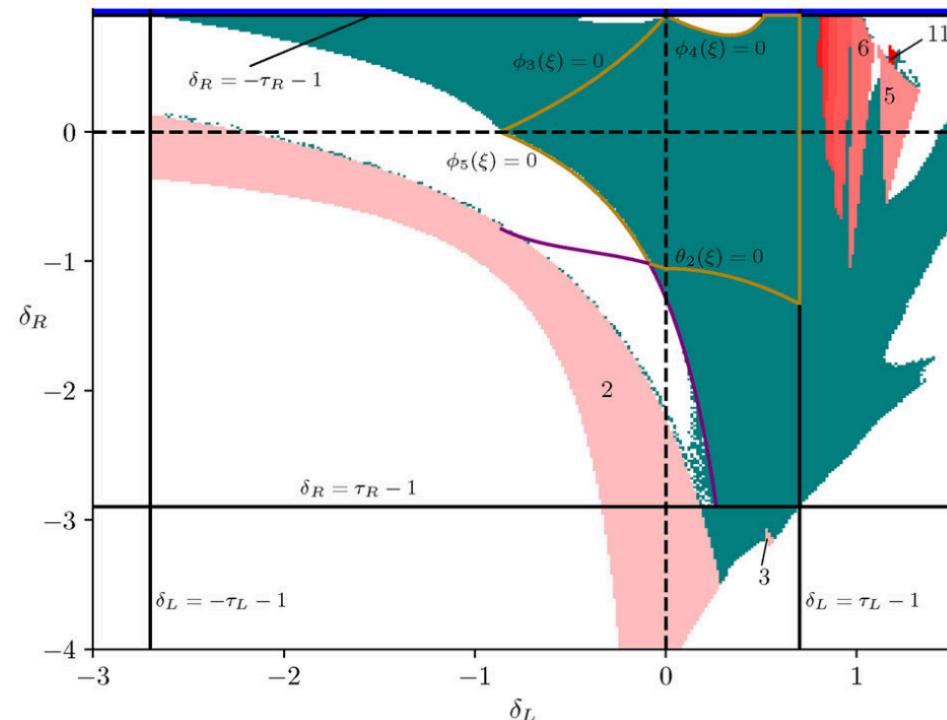


Figure: A 2D sketch of  $\Phi_{\text{trap}} \cap \Phi_{\text{cone}} \subset \mathbb{R}^4$

# The orientation reversing case

1. Let

$$\Phi^{(2)} = \{\xi \in \Phi \mid \delta_L < 0, \delta_R < 0\}$$

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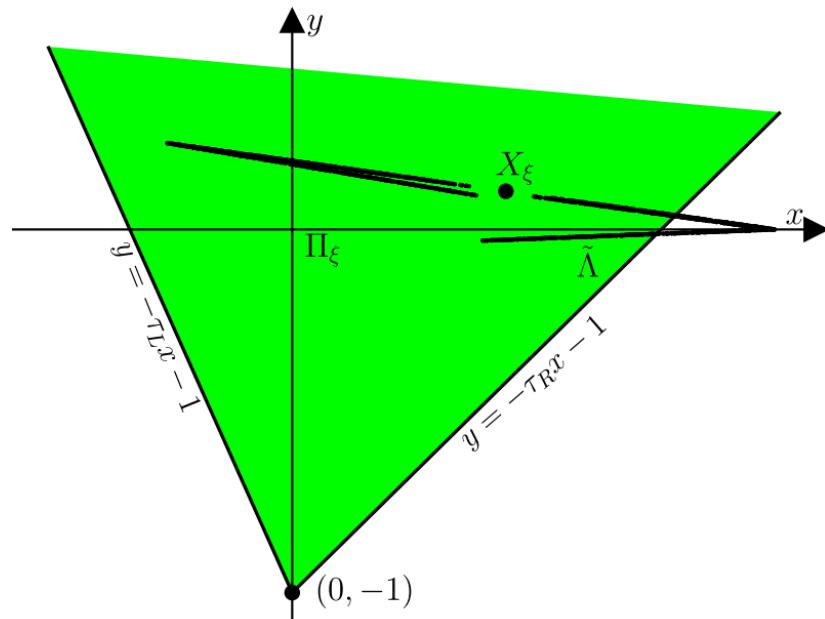
2. The attractor  $\Lambda$  which is again a closure of the unstable manifold of  $X$  has a crisis at  $\zeta_0^{(2)} = 0$  where

$$\zeta_0^{(2)} = \phi^-(\xi) = \delta_R - (\delta_R + \tau_R - (1 + \lambda_R^u)\lambda_L^u)\lambda_L^u.$$

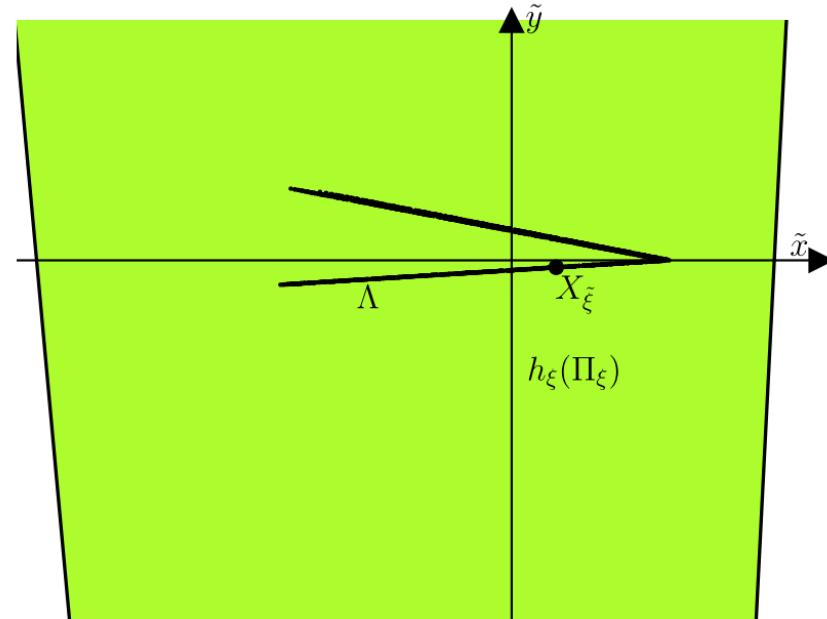
# The orientation reversing case

Proposition (Ghosh, McLachlan, & Simpson, 2024)

If  $\xi \in \mathcal{R}_n^{(2)}$  with  $n \geq 1$ , then  $g(\xi) \in \mathcal{R}_{n-1}^{(1)}$ .



(a)  $\xi = \xi_{\text{ex}}^{(2)} \in \mathcal{R}_1^{(2)}$

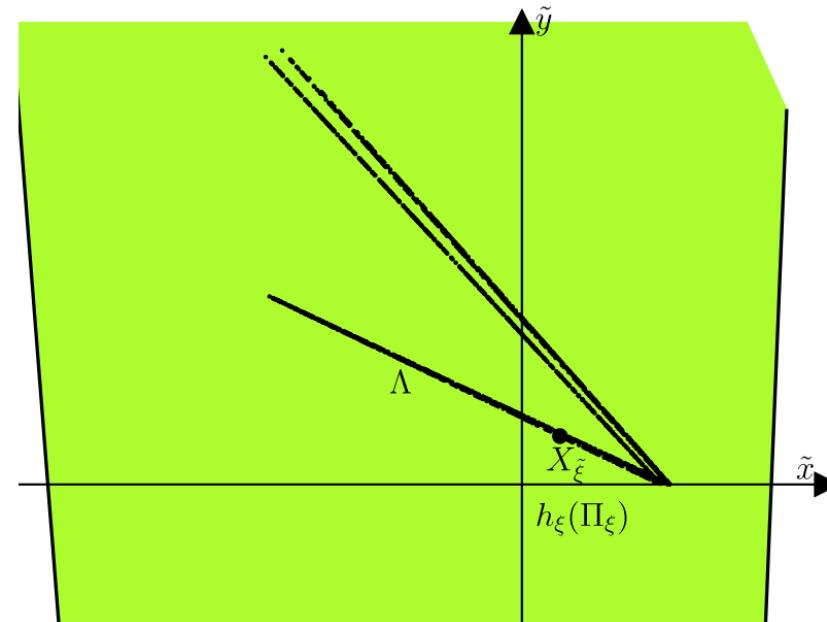
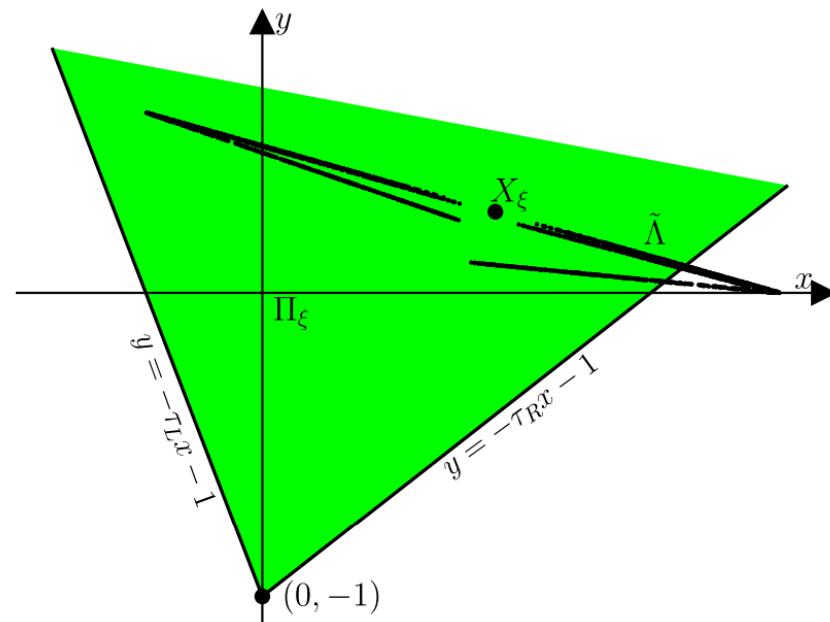


(b)  $\xi = g(\xi_{\text{ex}}^{(2)}) \in \mathcal{R}_0^{(1)}$

# The non-invertible case ( $\delta_L > 0, \delta_R < 0$ )

Proposition (Ghosh, McLachlan, & Simpson, 2024)

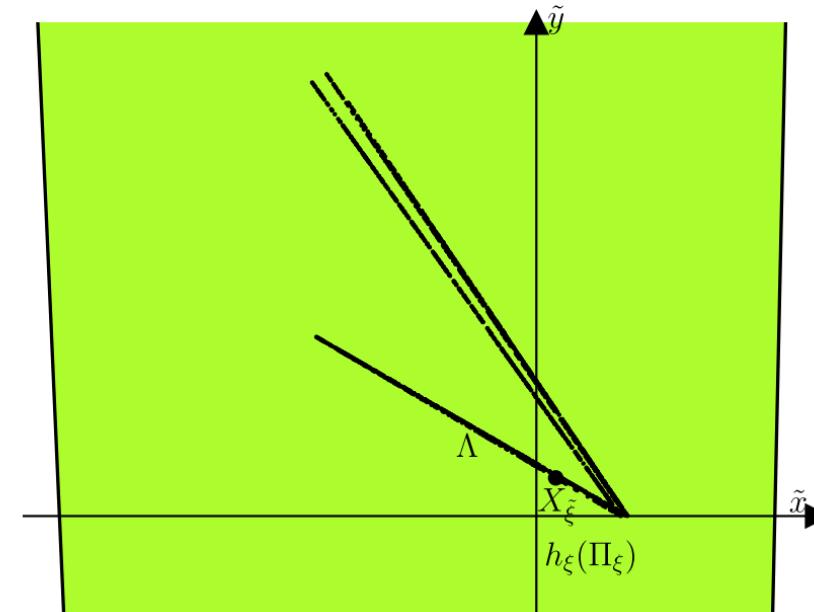
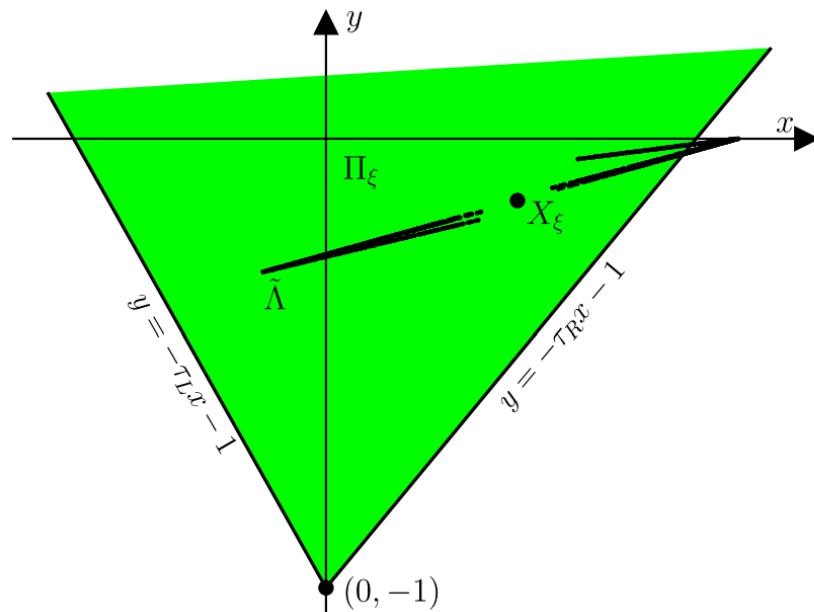
If  $\xi \in \mathcal{R}_n^{(3)}$  with  $n \geq 1$ , then  $g(\xi) \in \mathcal{R}_{n-1}^{(3)}$ .



# The non-invertible case ( $\delta_L < 0, \delta_R > 0$ )

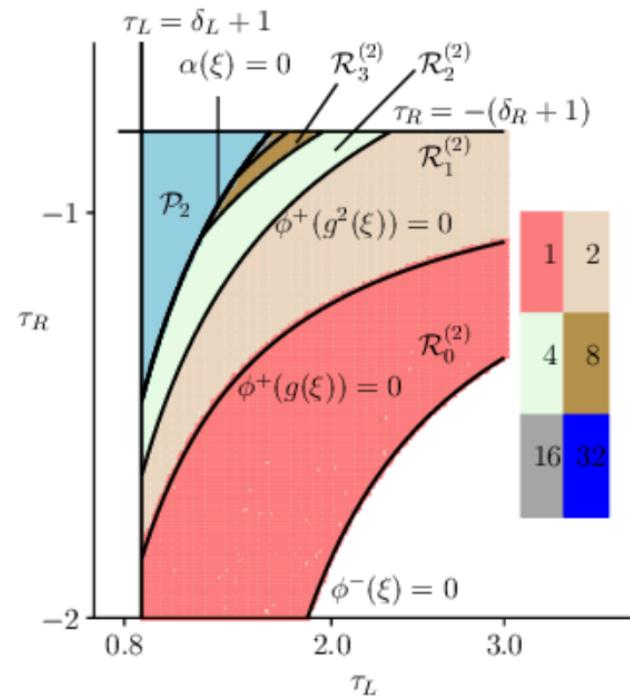
Proposition (Ghosh, McLachlan, & Simpson, 2024)

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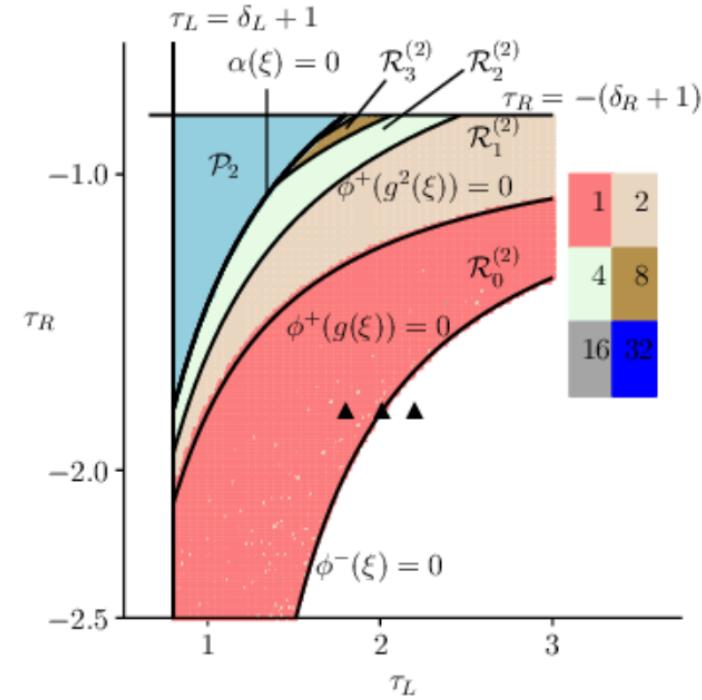


# Numerics

We verify these using Eckstein's greatest common divisor algorithm (2007), described by Avrutin *et al.*, 2007 to estimate from sample orbits the number of connected components in the attractor

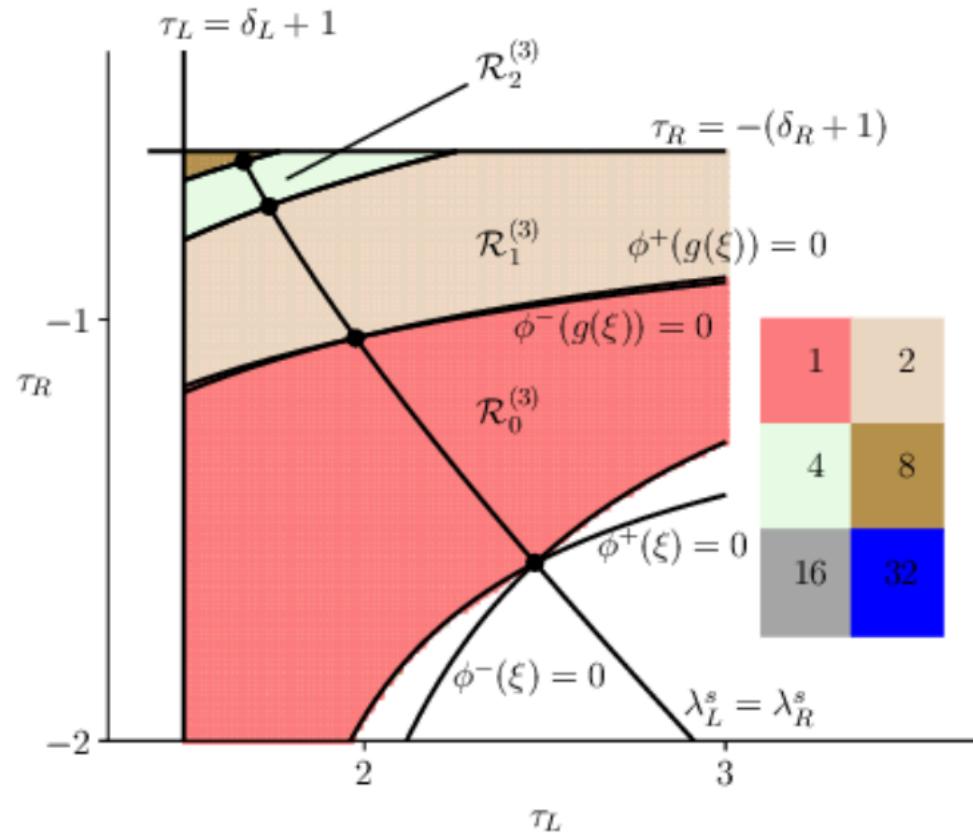
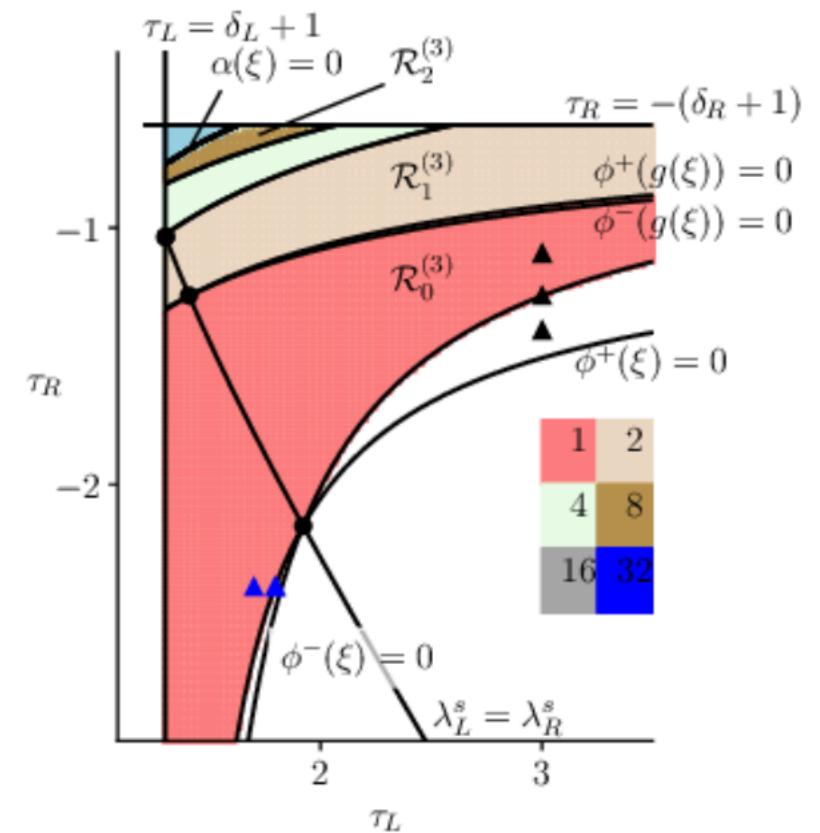


(a)  $\delta_L = -0.1, \delta_R = -0.2$ .

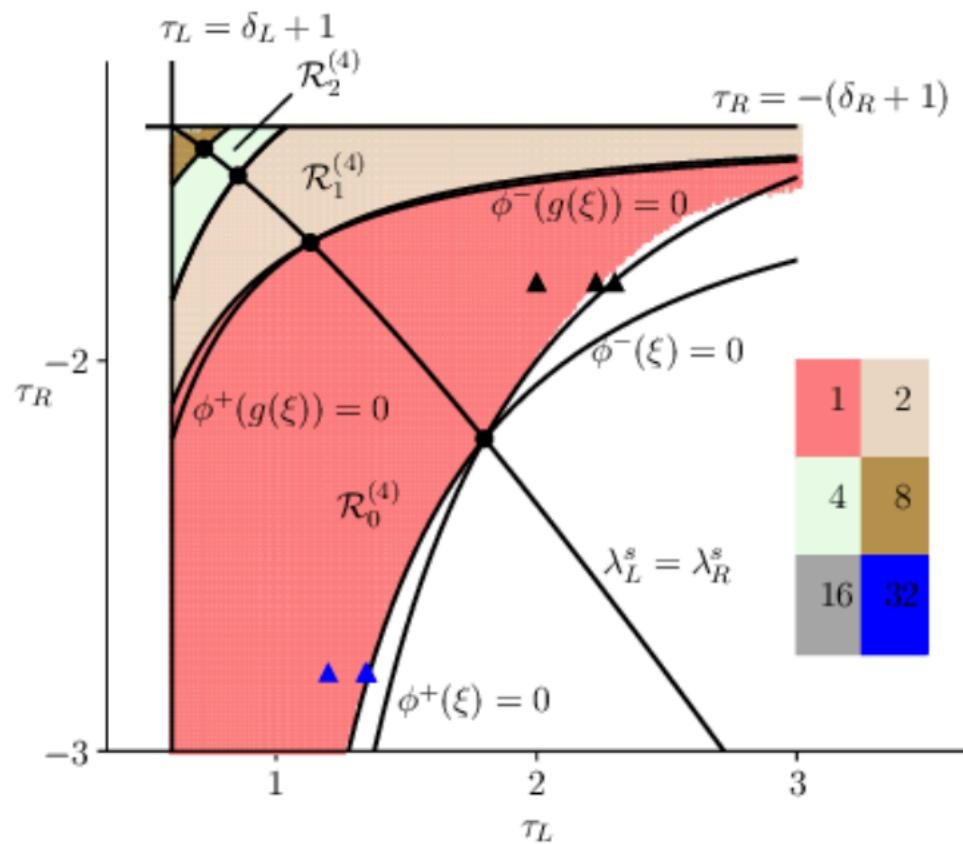
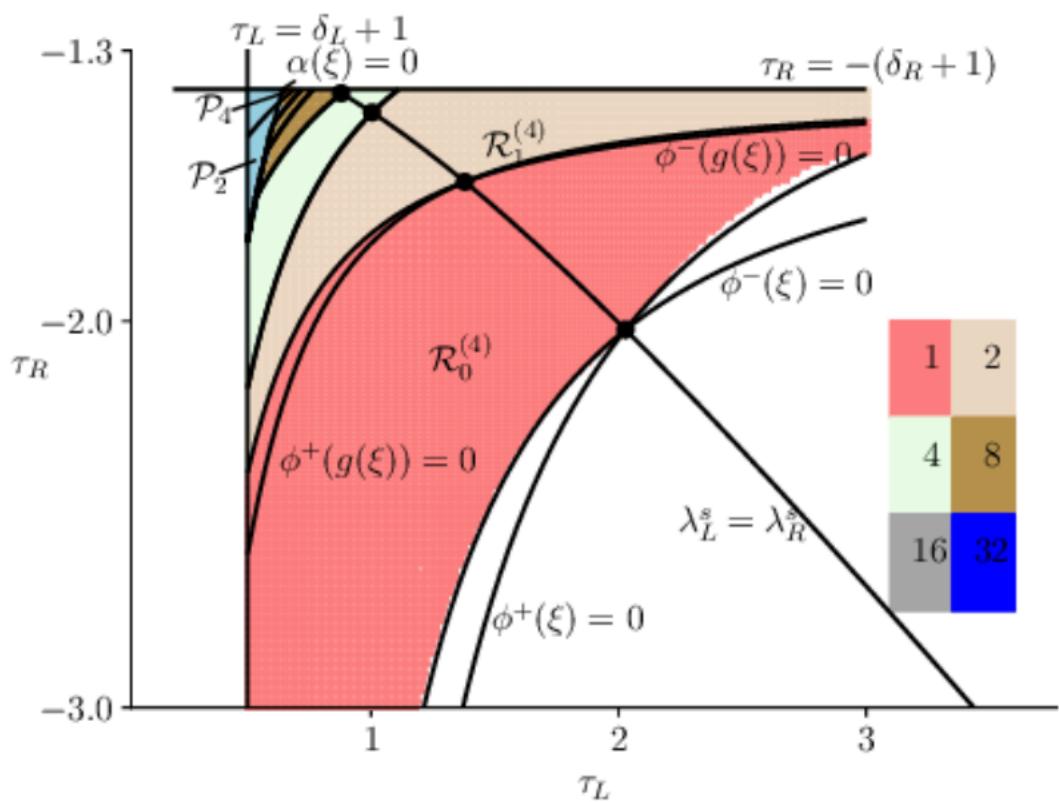


(b)  $\delta_L = -0.2, \delta_R = -0.2$ .

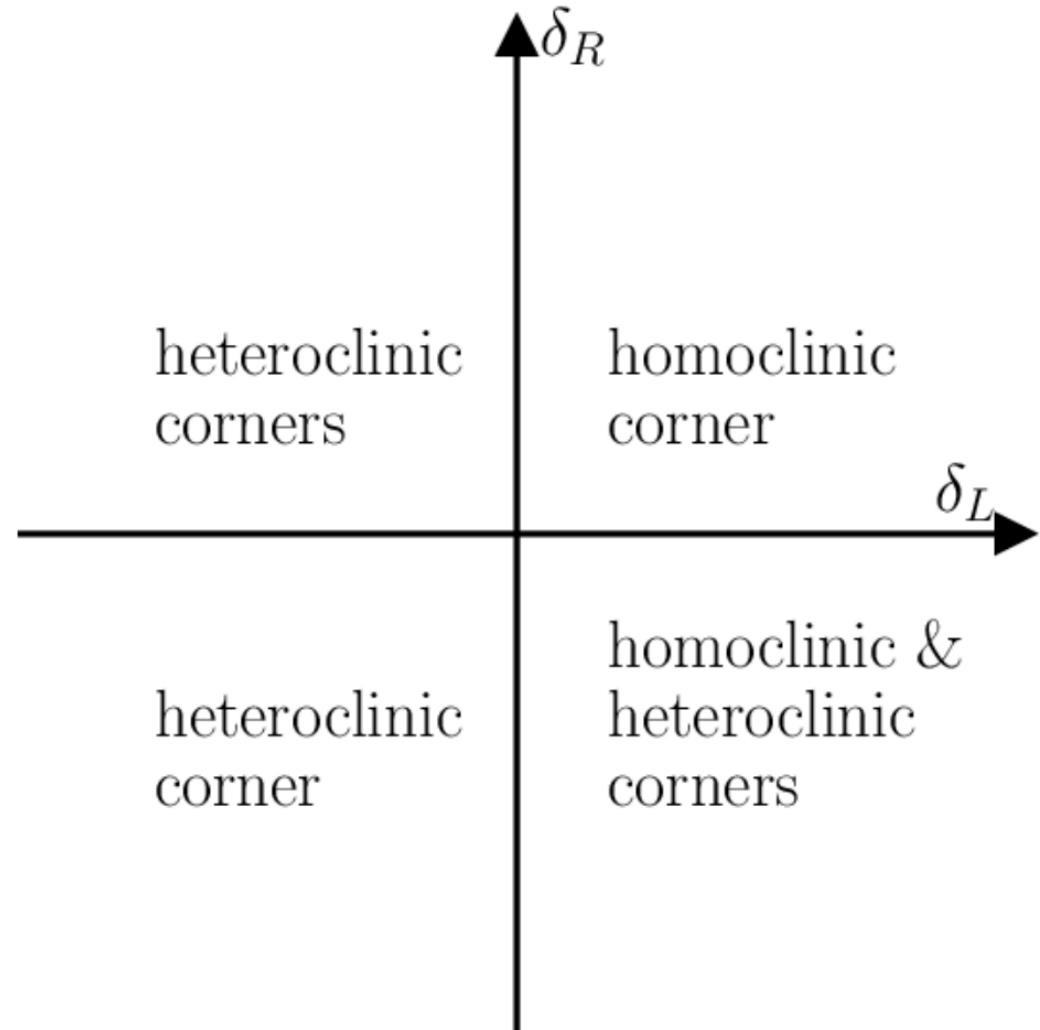
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(a)  $\delta_L = 0.5, \delta_R = -0.4.$ (b)  $\delta_L = 0.3, \delta_R = -0.4.$

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# Numerics



# Higher-dimensional setting

Let  $n \geq 2$ . Suppose  $\alpha > 1$  is an eigenvalue of  $A_L$ , and  $-\beta < -1$  of  $A_R$  with multiplicity one, and all other eigenvalues of  $A_L$  and  $A_R$  have modulus at most  $0 < r < 1$ .

Theorem (Ghosh, & Simpson, 2024)

*Holding the above assumption and*

$$r(n-1) < \frac{3}{7} \left( 1 - \frac{1}{\alpha} \right), \quad r(n-1) < \frac{3}{7} \left( 1 - \frac{1}{\beta} \right), \quad r(n-1) < \frac{1}{10} \left( \frac{1}{\alpha} + \frac{1}{\beta} - 1 \right),$$

*Then  $f$  has a topological attractor with a positive Lyapunov exponent.*

# Higher-dimensional setting

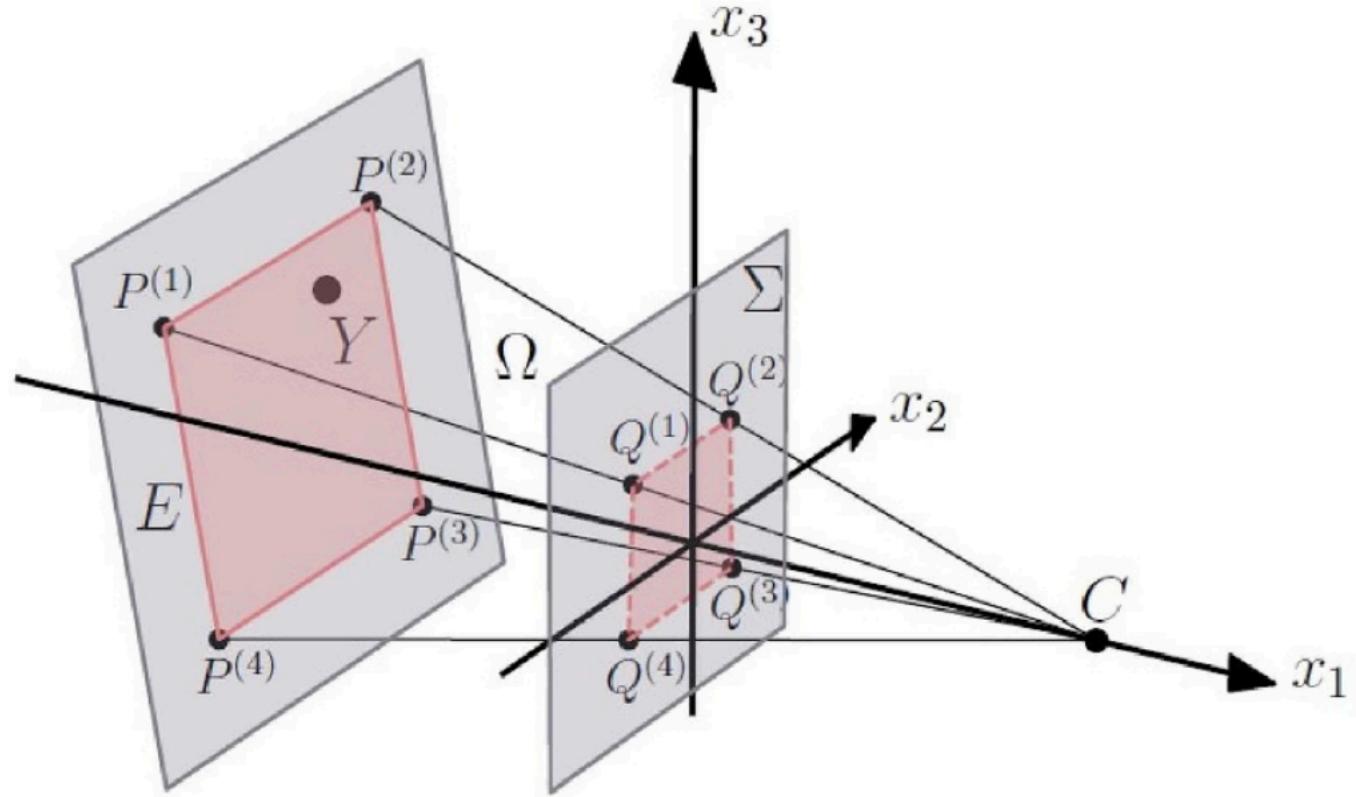


Figure: Construction of a forward invariant region  $\Omega$  for  $n = 3$

# Higher-dimensional setting

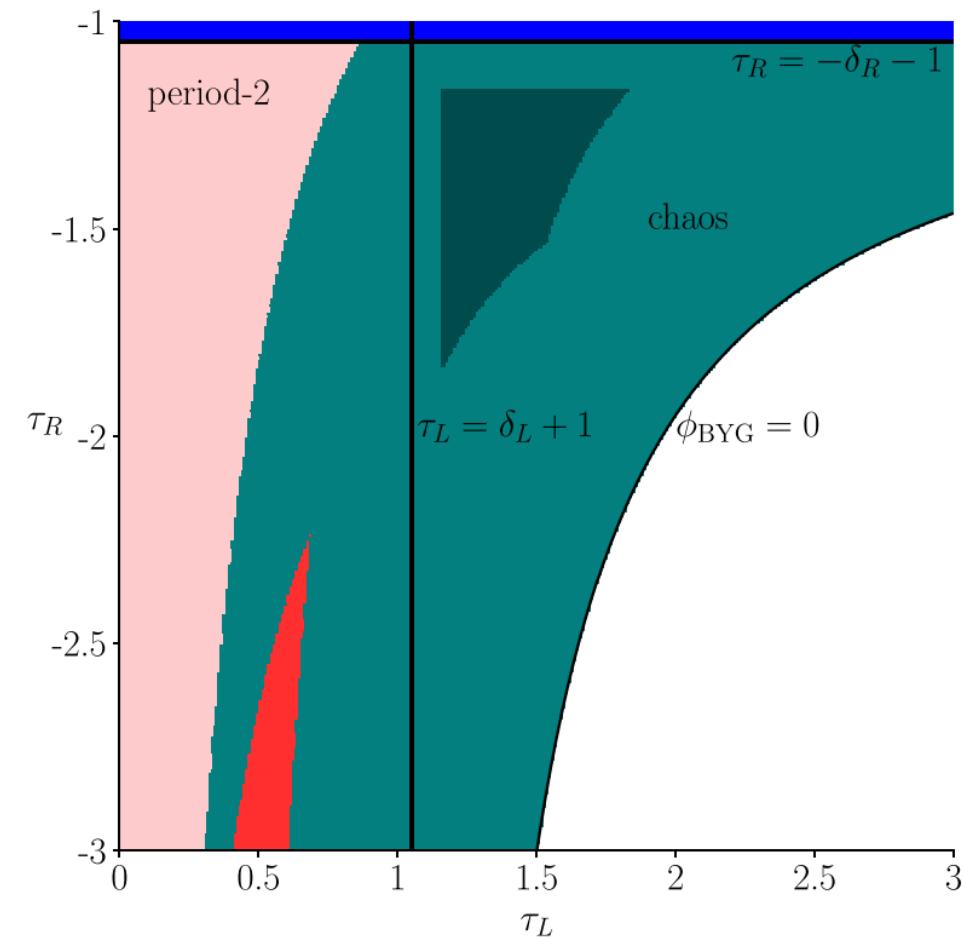


Figure: Robust chaos parameter region for the 2D map. We choose  $n = 2$  for simplicity

# Future directions

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3. Maps with multiple directions of instability is just as relevant: “**wild chaos**”.
4. Application: higher-dimensional construction as the key space for an encryption scheme.

# Acknowledgements



David Simpson



Robert McLachlan

MARSDEN FUND  
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A MARSDEN

ROYAL  
SOCIETY  
TE APĀRANGI

# Questions, comments, suggestions?

**Me understanding  
the 2-D BCNF**



$(\tau_l, \delta_l, \tau_r, \delta_r)$  behaving nicely

**Renormalisation  
Let's do that again...  
but worse**



wild  
chaos

$\Omega$  JYC

0 11...