

## **Dimensões generalizadas e a conjectura de Kaplan-Yorke**

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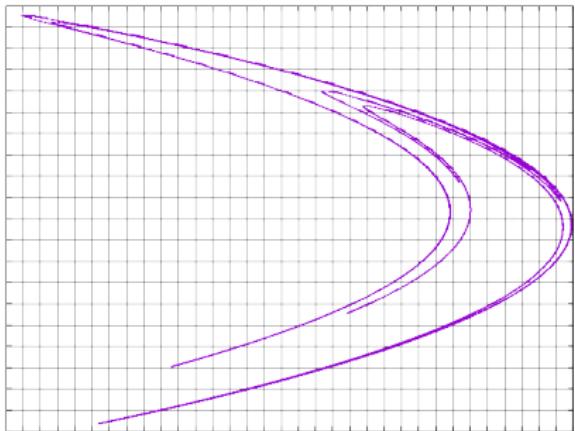
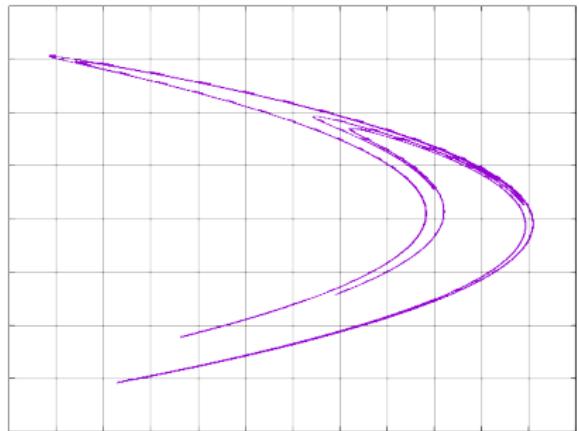
Disciplina: Caos em Sistemas Dissipativos

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# Objetivos

- Introduzir o conceito de dimensão generalizada
- Expor as dimensões mais importantes, a saber,  $D_0$ ,  $D_1$  e  $D_2$
- Apresentar a conjectura de Kaplan-Yorke
- Discutir a relação entre geometria e dinâmica



## Dimensão da contagem de caixas

$$D = \lim_{\epsilon \rightarrow 0} \frac{\ln N(\epsilon)}{\ln(1/\epsilon)}$$

## Medida natural

$$\mu_i = \lim_{T \rightarrow \infty} \frac{\eta(C_i, \mathbf{x}_0, T)}{T}$$

$\eta(C_i, \mathbf{x}_0, T)$ : tempo que a órbita que se origina em  $\mathbf{x}_0$  passa na caixa  $C_i$  durante o intervalo de tempo  $0 \leq t \leq T$ .

## Espectro de dimensões generalizadas

$$D_q = \frac{1}{1-q} \lim_{\epsilon \rightarrow 0} \frac{\ln I(q, \epsilon)}{\ln(1/\epsilon)}, \quad I(q, \epsilon) = \sum_{i=1}^{N(\epsilon)} \mu_i^q$$

$$D_{q_1} \leq D_{q_2} \text{ se } q_1 > q_2$$

## Dimensão $D_0$

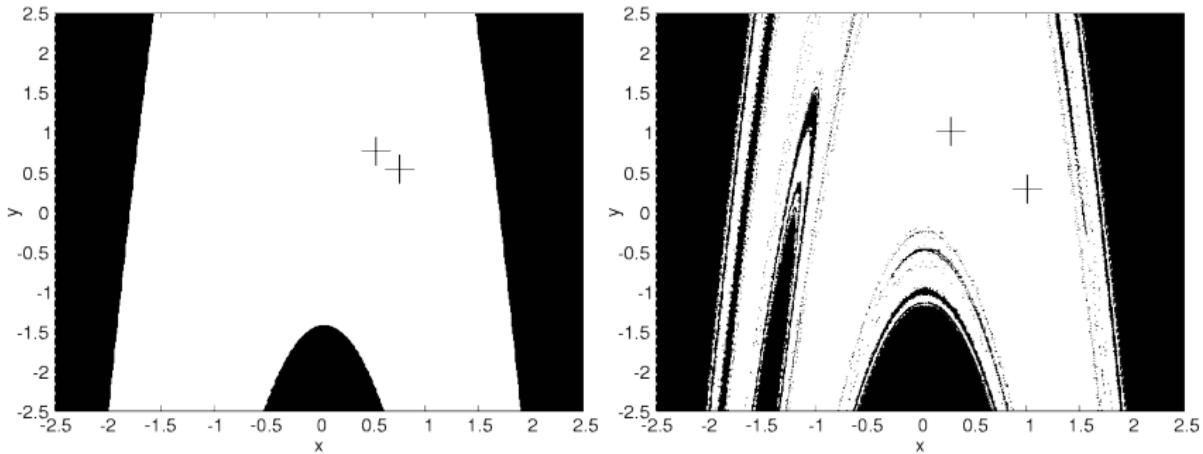
- $q = 0 \Rightarrow I(0, \epsilon) = \sum_{i=1}^{N(\epsilon)} \mu_i^0 = N(\epsilon)$

$\therefore D_0 = \lim_{\epsilon \rightarrow 0} \frac{\ln N(\epsilon)}{\ln(1/\epsilon)} \Rightarrow D_0$  é a dimensão da contagem de caixas

- $\mu_i = \frac{1}{N(\epsilon)} \Rightarrow I(q, \epsilon) = \sum_{i=1}^{N(\epsilon)} \left( \frac{1}{N(\epsilon)} \right)^q = N(\epsilon)^{1-q}$

$\therefore D_q = D_0 , \forall q$

# Importância de $D_0$



## Função incerteza

$$f(\epsilon) \sim \epsilon^{2-D_0}$$

# Dimensão $D_1$

Definimos

$$D_1 = \lim_{q \rightarrow 1} D_q = \lim_{\epsilon \rightarrow 0} \lim_{q \rightarrow 1} \frac{1}{1-q} \frac{\ln I(q, \epsilon)}{\ln(1/\epsilon)}, \quad I(q, \epsilon) = \sum_{i=1}^{N(\epsilon)} \mu_i^q$$

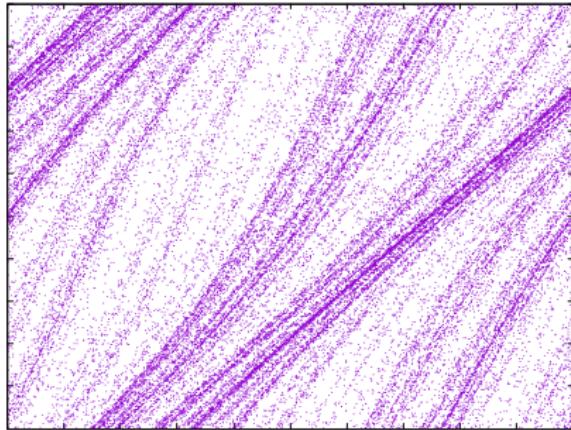
Dimensão de informação

$$D_1 = \lim_{\epsilon \rightarrow 0} \frac{\sum_{i=1}^{N(\epsilon)} \mu_i \ln \mu_i}{\ln \epsilon}$$

# Importância de $D_1$

## Mapa de Sinai

$$\begin{aligned}x_{n+1} &= x_n + y_n + \delta \cos(2\pi y_n) \quad \text{mod } 1 \\y_{n+1} &= x_n + 2y_n \quad \text{mod } 1\end{aligned}$$



$$\begin{aligned}D_0 &= 2 \\D_1 &< 2\end{aligned}$$

$$D_0(\theta) = D_1, \quad 0 < \theta < 1$$

## Dimensão $D_2$

“Integral” de correlação

$$C(\epsilon) = \lim_{K \rightarrow \infty} \frac{1}{K^2} \sum_{ij}^K \Theta(\epsilon - |\mathbf{x}_i - \mathbf{x}_j|)$$

Podemos mostrar que

$$C(\epsilon) \sim I(2, \epsilon) = \sum_{i=1}^{N(\epsilon)} \mu_i^2$$

Portanto

$$D_2 = \lim_{\epsilon \rightarrow 0} \frac{\ln C(\epsilon)}{\ln \epsilon}$$

# Importância de $D_2$

- Dados experimentais
- Alta dimensionalidade

Outro exemplo: periodicidade induzida em computadores

$$\bar{m} \sim \delta^{-D_2/2}$$

$\bar{m}$ : tamanho do período

$\delta$ : erro de arredondamento

# Conjectura de Kaplan-Yorke

## Dimensão de Lyapunov

$$D_L = K + \frac{1}{|h_{k+1}|} \sum_{j=1}^k h_j, \quad \sum_{j=1}^k h_j \geq 0$$

## Conjectura de Kaplan-Yorke

$$D_L = D_1$$

# Mapa do padeiro

## Definição geral

$$\mathcal{B}(x_n, y_n) = \begin{cases} \begin{pmatrix} \lambda_a & 0 \\ 0 & \frac{1}{\alpha} \end{pmatrix} \begin{pmatrix} x_n \\ y_n \end{pmatrix} & , \text{se } y < \alpha \\ \begin{pmatrix} \lambda_b & 0 \\ 0 & \frac{1}{\beta} \end{pmatrix} \begin{pmatrix} x_n \\ y_n \end{pmatrix} + \begin{pmatrix} 1 - \lambda_b \\ -\frac{\alpha}{\beta} \end{pmatrix} & , \text{se } y > \alpha \end{cases}$$

$$h_1 = \lim_{n \rightarrow \infty} \frac{1}{n} \ln \left[ \left( \frac{1}{\alpha} \right)^{n_1} \left( \frac{1}{\beta} \right)^{n_2} \right] = \alpha \ln \frac{1}{\alpha} + \beta \ln \frac{1}{\beta}$$

$$h_2 = \lim_{n \rightarrow \infty} \frac{1}{n} \ln [(\lambda_a)^{n_1} (\lambda_b)^{n_2}] = \alpha \ln \lambda_a + \beta \ln \lambda_b$$

# Mapa do padeiro

Por outro lado,

$$\hat{I}(q, \epsilon) = \hat{I}_a(q, \epsilon) + \hat{I}_b(q, \epsilon)$$

$$\hat{I}_a(q, \epsilon) = \alpha^q \hat{I}_a(q, \epsilon/\lambda_a)$$

$$\hat{I}_b(q, \epsilon) = \beta^q \hat{I}_b(q, \epsilon/\lambda_b)$$

$$\hat{I}(q, \epsilon) \simeq K \epsilon^{(q-1)\hat{D}_q}$$

## Equação transcendental para o mapa do padeiro

$$\alpha^q \lambda_a^{(q-1)\hat{D}_q} + \beta^q \lambda_b^{(q-1)\hat{D}_q} = 1$$

## Conjectura de Kaplan-Yorke satisfeita

$$D_L = 1 + \frac{\alpha \ln(1/\alpha) + \beta \ln(1/\beta)}{\alpha \ln(1/\lambda_a) + \beta \ln(1/\lambda_b)} = D_1$$

## Referências

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