#### Instituto de Física da USP

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

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Disciplina: Caos em Sistemas Dissipativos

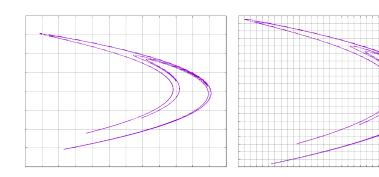
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# **Objetivos**

- Introduzir o conceito de dimensão generalizada
- Expor as dimensões mais importantes, a saber, D<sub>0</sub>, D<sub>1</sub> e D<sub>2</sub>
- Apresentar a conjectura de Kaplan-Yorke
- Discutir a relação entre geometria e dinâmica



#### Dimensão da contagem de caixas

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



#### Medida natural

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

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

### Espectro de dimensões generalizadas

$$D_q = rac{1}{1-q} \lim_{\epsilon o 0} rac{\ln I(q,\epsilon)}{\ln(1/\epsilon)}, \qquad I(q,\epsilon) = \sum_{i=1}^{N(\epsilon)} \mu_i^q$$

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



# Dimensão D<sub>0</sub>

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

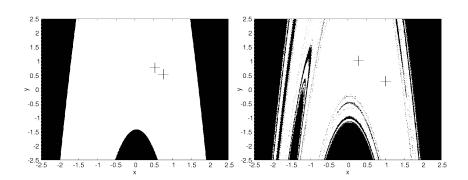
$$\therefore D_0 = \lim_{\epsilon \to 0} \frac{\ln N(\epsilon)}{\ln(1/\epsilon)} \Rightarrow D_0$$
 é a dimensão da contagem de caixas

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

$$\therefore D_q = D_0 , \forall q$$



# Importância de D<sub>0</sub>



### Função incerteza

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

## Dimensão D<sub>1</sub>

#### **Definimos**

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

### Dimensão de informação

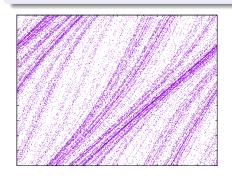
$$D_1 = \lim_{\epsilon \to 0} \frac{\sum_{i=1}^{N(\epsilon)} \mu_i \ln \mu_i}{\ln \epsilon}$$



# Importância de D<sub>1</sub>

### Mapa de Sinai

$$x_{n+1} = x_n + y_n + \delta \cos(2\pi y_n) \mod 1$$
  
 $y_{n+1} = x_n + 2y_n \mod 1$ 



$$D_0 = 2$$
  
 $D_1 < 2$ 

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



## Dimensão D<sub>2</sub>

#### "Integral" de correlação

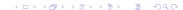
$$C(\epsilon) = \lim_{K o \infty} rac{1}{K^2} \sum_{ij}^K \Theta(\epsilon - |oldsymbol{x}_i - oldsymbol{x}_j|)$$

#### Podemos mostrar que

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

#### **Portanto**

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



# Importância de D<sub>2</sub>

- 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

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

### Conjectura de Kaplan-Yorke

$$D_L = D_1$$

# Mapa do padeiro

### Definição geral

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

$$h_{1} = \lim_{n \to \infty} \frac{1}{n} \ln \left[ \left( \frac{1}{\alpha} \right)^{n_{1}} \left( \frac{1}{\beta} \right)^{n_{2}} \right] = \alpha \ln \frac{1}{\alpha} + \beta \ln \frac{1}{\beta}$$

$$h_{2} = \lim_{n \to \infty} \frac{1}{n} \ln \left[ \left( \lambda_{a} \right)^{n_{1}} \left( \lambda_{b} \right)^{n_{2}} \right] = \alpha \ln \lambda_{a} + \beta \ln \lambda_{b}$$



## Mapa do padeiro

Por outro lado,

$$\hat{l}(q,\epsilon) = \hat{l}_a(q,\epsilon) + \hat{l}_b(q,\epsilon)$$
 $\hat{l}_a(q,\epsilon) = \alpha^q \hat{l}_a(q,\epsilon/\lambda_a)$ 
 $\hat{l}_b(q,\epsilon) = \beta^q \hat{l}_b(q,\epsilon/\lambda_b)$ 
 $\hat{l}(q,\epsilon) \simeq K\epsilon^{(q-1)\hat{D}_q}$ 

## 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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