# Mechanisms of Chaos: Stable Instability

Reading for this lecture:

NDAC, Sec. 12.0-12.3, 9.3, and 10.5.

## Unpredictability:

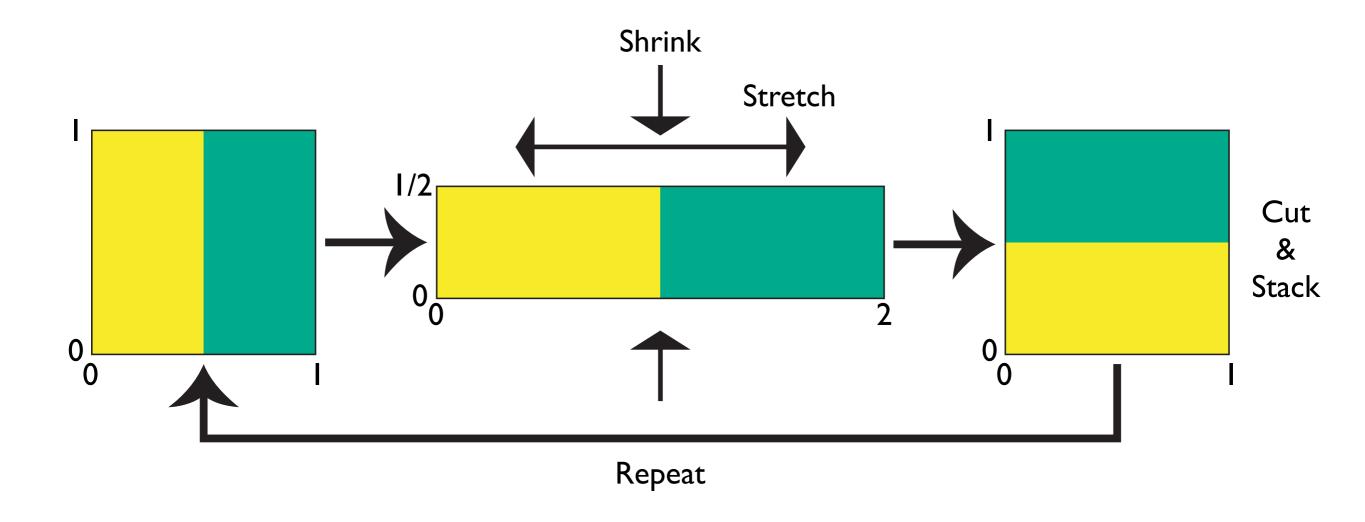
Orbit complicated: difficult to follow Repeatedly convergent and divergent Net amplification of small variations

What geometry produces this?

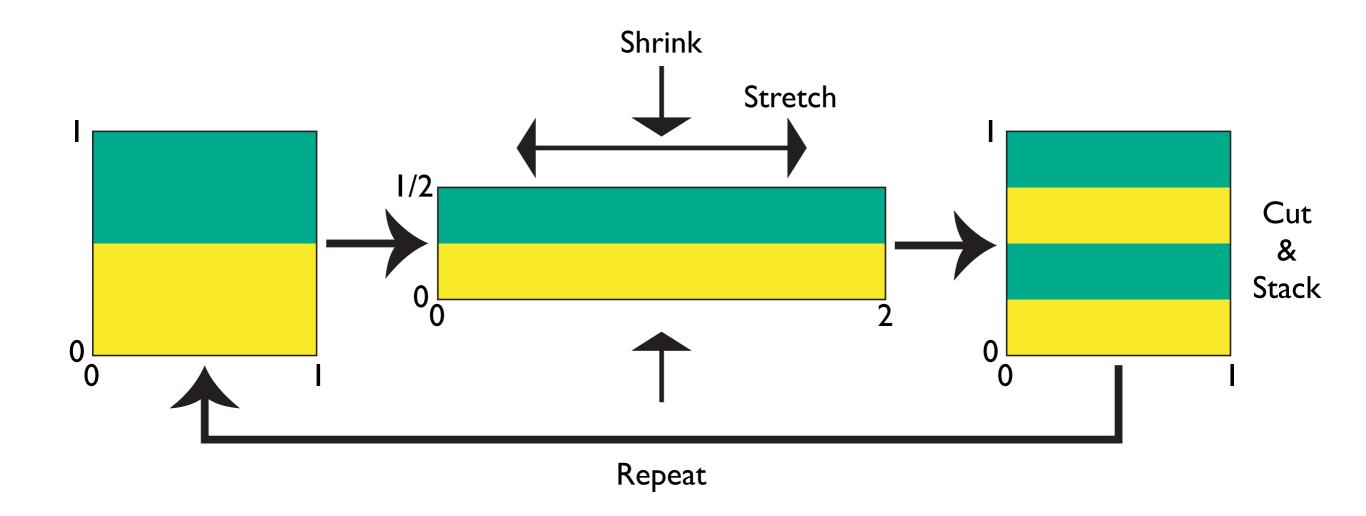
#### Stretch and fold:

Flow stretches state space
But to be stable, must be done in a compact region
Must fold back into region

# Baker's transformation: kneading state space



# Baker's transformation ... kneading state space



Baker's transformation ...

# 2D Baker's Map:

$$(x_n, y_n) \in [0, 1] \times [0, 1]$$

$$x_{n+1} = 2x_n \pmod{1}$$

$$y_{n+1} = \begin{cases} \frac{1}{2}y_n, & x_n \leq \frac{1}{2} \\ \frac{1}{2} + \frac{1}{2}y_n, & x_n > \frac{1}{2} \end{cases}$$

Baker's transformation ...

$$A = \begin{pmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{pmatrix}$$

Calculate:

$$\lambda_1 = 2$$

$$\lambda_1=2$$
 Stretch  $\vec{v}_1=(1,0)$  Only horizontal

$$\lambda_2 = 1/2$$

$$\lambda_2=1/2$$
 Shrink  $\vec{v}_2=(0,1)$  Only vertical

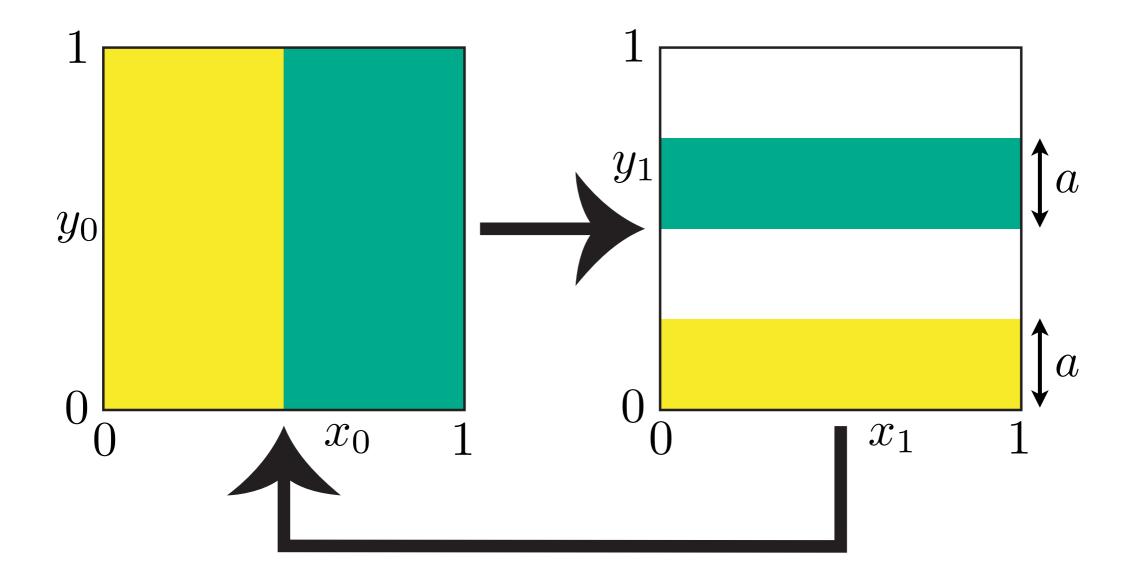
$$Tr(A) = 5/2$$

$$Det(A) = 1$$

 $\mathrm{Det}(A) = 1$  Area preserving: No attractor per se

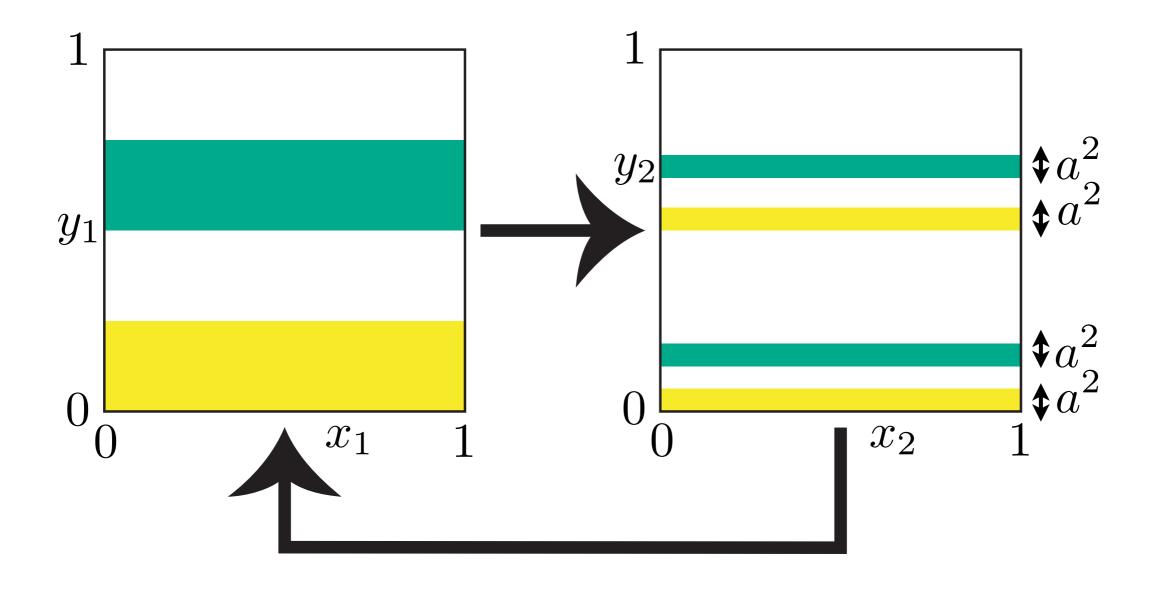
Independent of  $\vec{x}$ 

# Dissipative Baker's Map:



Mechanisms of Chaos ...

Dissipative Baker's Map ... again!



Dissipative Baker's Map ...

$$x_{n+1} = 2x_n \pmod{1}$$

$$y_{n+1} = \begin{cases} ay_n, & x_n \le \frac{1}{2} \\ \frac{1}{2} + ay_n, & x_n > \frac{1}{2} \end{cases}$$

$$a \in \left[0, \frac{1}{2}\right]$$

Dissipative Baker's Map ...

Stability? 
$$A = \begin{pmatrix} 2 & 0 \\ 0 & a \end{pmatrix}$$

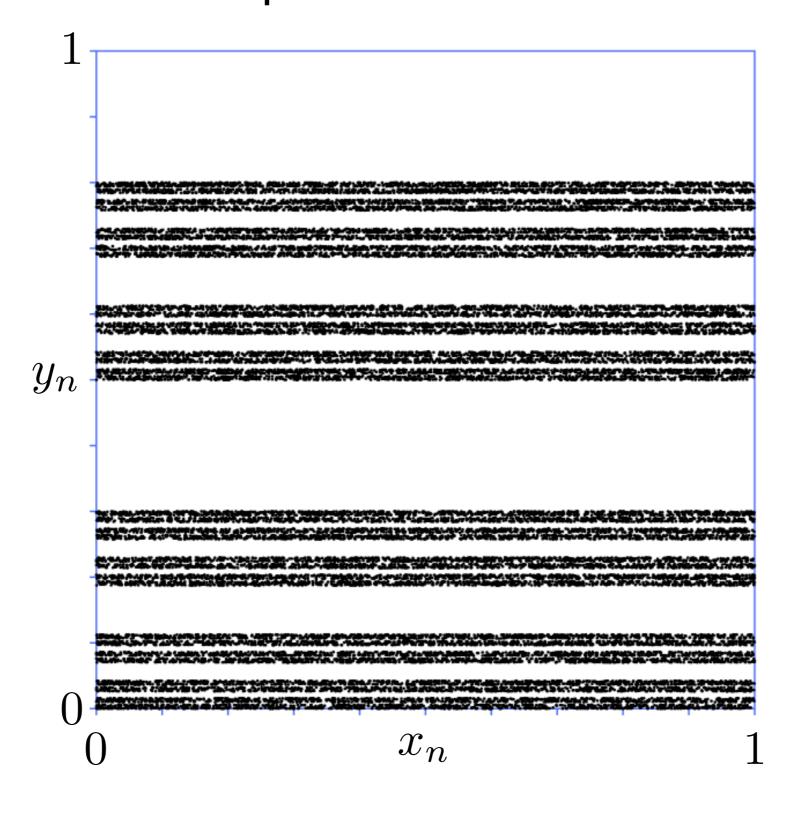
#### Calculate:

$$\lambda_1=2$$
  $\vec{v}_1=(1,0)$   $\lambda_2=a$   $\vec{v}_2=(0,1)$  Independent of  $\vec{x}$ 

$$\mathrm{Det}(A) = 2a$$
 Dissipative:  $a < 1/2$  Area contraction

Attractor!

# Dissipative Baker's Map Simulation: a = 0.3



Dissipative Baker's Map ...

Stability? (x,y) versus  $(x+\epsilon,y+\delta)$ 

$$\Delta x_1 = 2(x_0 + \epsilon) - 2x_0 = 2\epsilon$$

$$\Delta y_1 = a(y_0 + \delta) - ay_0 = a\delta$$

$$\Delta x_n = 2^n \epsilon$$
 Exponential Growth of Errors

$$\Delta y_n = a^n \delta$$
 Exponential Stability

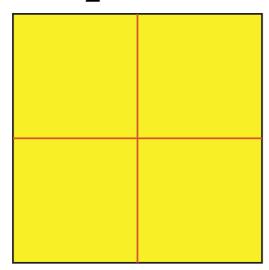
#### Dimension of a Set:

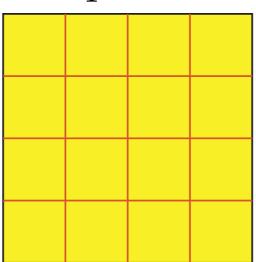
Number of boxes to cover set at given measurement resolution:

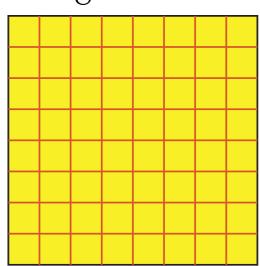
$$\epsilon = \frac{1}{2}$$
  $N = 4$ 

$$\epsilon = \frac{1}{4} \ N = 16$$

$$\epsilon = \frac{1}{2} \quad N = 4$$
  $\epsilon = \frac{1}{4} \quad N = 16$   $\epsilon = \frac{1}{8} \quad N = 64$ 







$$N(\epsilon = \frac{1}{2^n}) = \left(\frac{1}{2^n}\right)^{-2} = 2^{2n}$$
$$N(\epsilon) \propto \epsilon^{-2}$$

## Generalizing

$$N(\epsilon) \propto \epsilon^{-d}$$

Or (Definition) dimension: 
$$d = \lim_{\epsilon \to 0} -\frac{\log N(\epsilon)}{\log \epsilon}$$

Dimension of Dissipative Baker's Attractor ...

At iteration n:

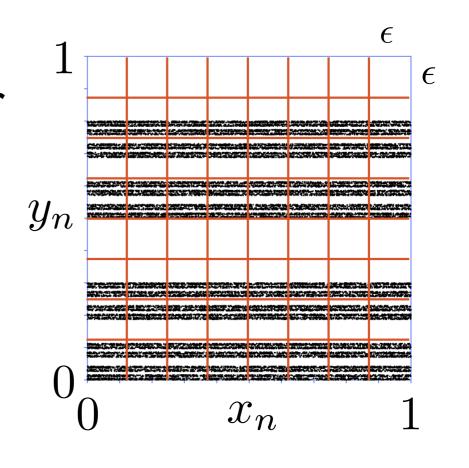
 $2^n$  strips of thickness  $a^n$ 

How many boxes  $N(\epsilon)$  to cover attractor at resolution  $\epsilon$ ?

Take: 
$$\epsilon = a^n$$

Number of boxes for each strip:  $a^{-n}$ 

$$N(\epsilon) = a^{-n} \times 2^n = \left(\frac{a}{2}\right)^{-n}$$



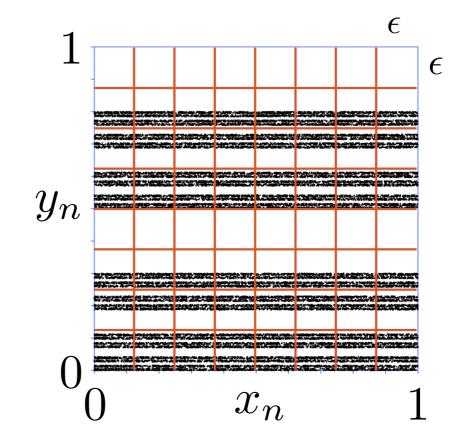
# Dimension of Dissipative Baker's Attractor ...

#### Dimension:

$$d = \lim_{\epsilon \to 0} -\frac{\log N(\epsilon)}{\log \epsilon}$$

$$= \lim_{n \to \infty} -\frac{\log(a/2)^{-n}}{\log a^n}$$

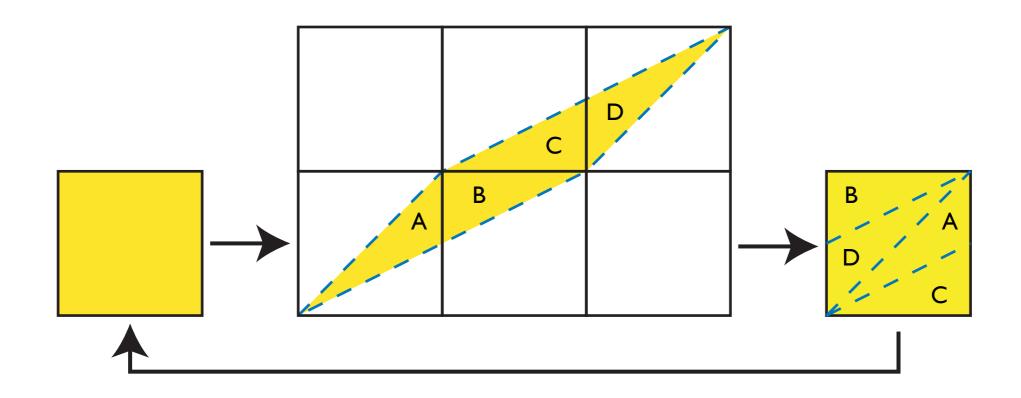
$$= 1 + \frac{\log \frac{1}{2}}{\log a}$$



$$a = 0.3 \Rightarrow d = 1.576... < 2$$
!

Area preserving: as  $a \to \frac{1}{2}, d \to 2$ 

# Cat map (aka toral automorphism): $(x, y) \in \mathbf{T}^2$ Intrinsic stretch/shrink directions



$$\begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_n \\ y_n \end{pmatrix} \pmod{1}$$

Fixed point:  $\vec{x}^* = (0,0)$ 

# Cat map (aka Toral automorphism) ...

$$\begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_n \\ y_n \end{pmatrix} \pmod{1}$$

Calculate: 
$$\lambda_1 =$$

$$\lambda_1 = rac{3+\sqrt{5}}{2} > 1$$
 stretch

$$\begin{array}{ll} \textbf{Calculate:} & \lambda_1=\frac{3+\sqrt{5}}{2}>1 \quad \text{stretch} & \vec{v}_1=(\frac{1+\sqrt{5}}{2},1) \\ & \lambda_2=\frac{3-\sqrt{5}}{2}<1 \quad \text{shrink} & \vec{v}_2=(\frac{1-\sqrt{5}}{2},1) \end{array}$$

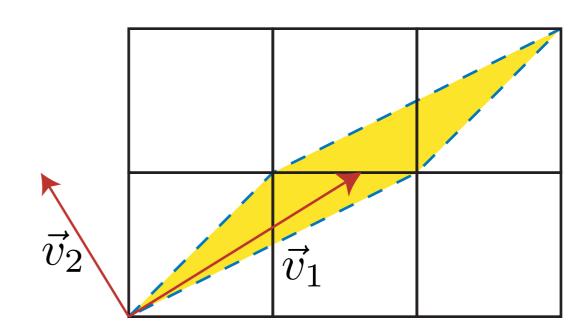
$$\vec{v}_1 = (\frac{1+\sqrt{5}}{2}, 1)$$

$$\vec{v}_2 = (\frac{1 - \sqrt{5}}{2}, 1)$$

$$Tr(A) = 3$$

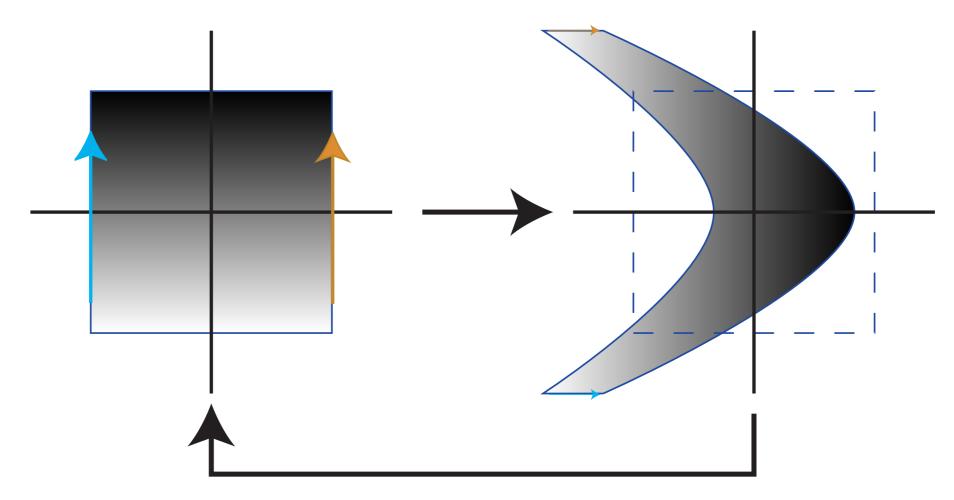
$$\mathrm{Det}(A) = 1$$
 area preserving

# Independent of $\vec{x}$





Hénon map: 
$$(x,y) \in \mathbf{R}^2$$
 
$$x_{n+1} = y_n + 1 - ax_n^2$$
 
$$y_{n+1} = bx_n$$



Stretch and fold depend on location

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Hénon map ...

Stretch & fold depend on location:

Jacobian:

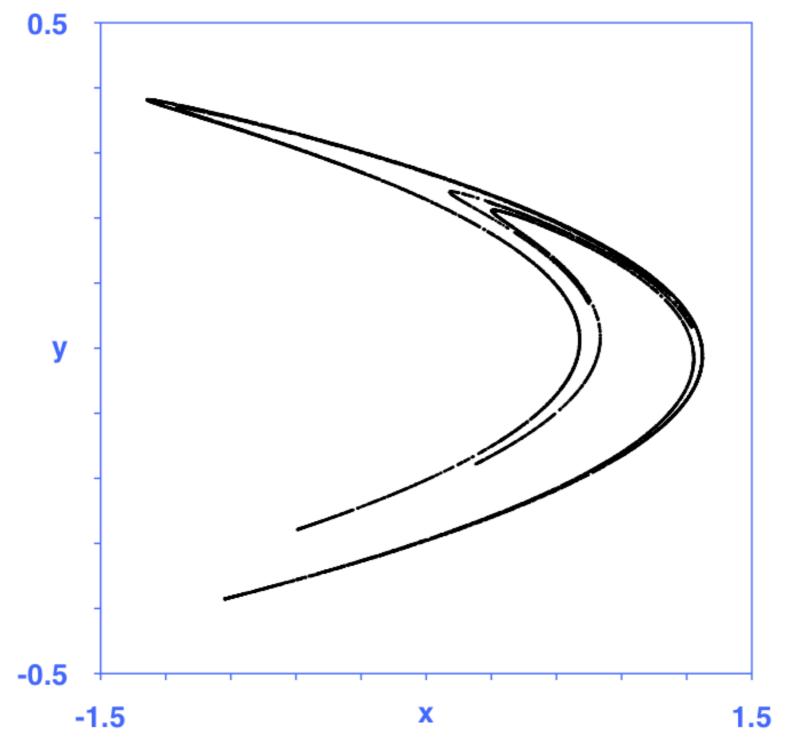
$$A = \begin{pmatrix} -2ax_n & 1\\ b & 0 \end{pmatrix}$$

Dissipative (and orientation reversing):

$$Det(A) = -b$$

#### Hénon Attractor:

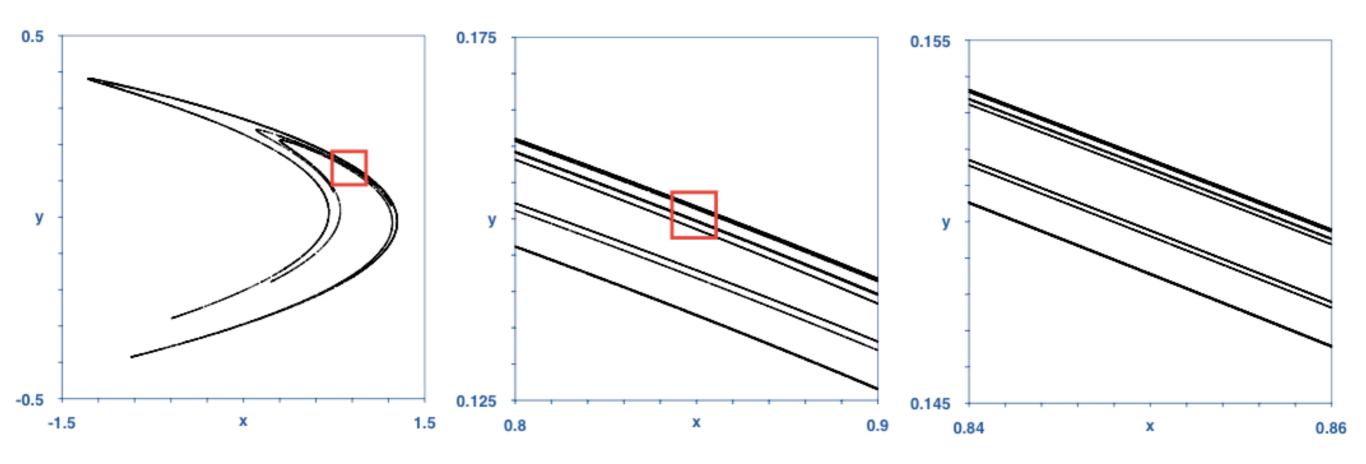
Control parameters: (a, b) = (1.4, 0.3)



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# Mechanisms of Chaos ... Henon Attractor ...

#### Self-similar:

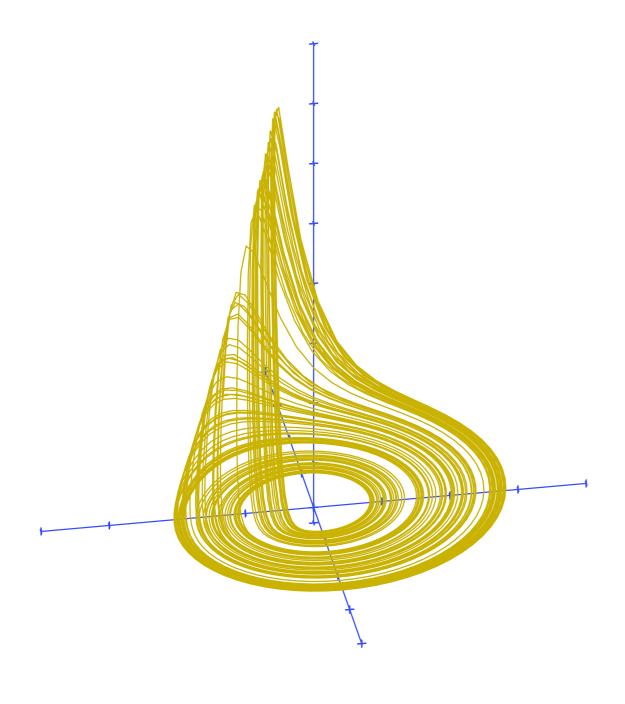


# Self-similar attractor = Dissipation + Instability

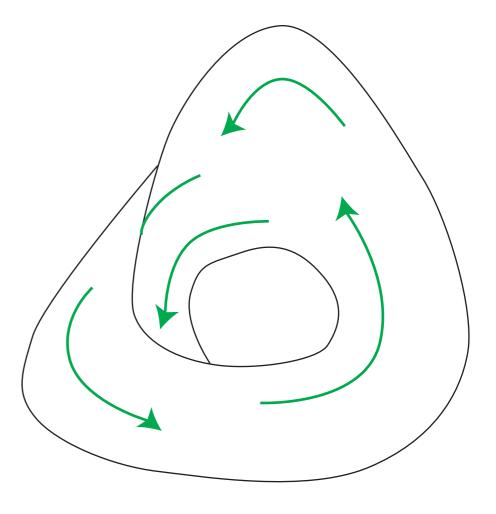
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Mechanisms of Chaos ... How does the stretch & fold mechanism work in ODEs?

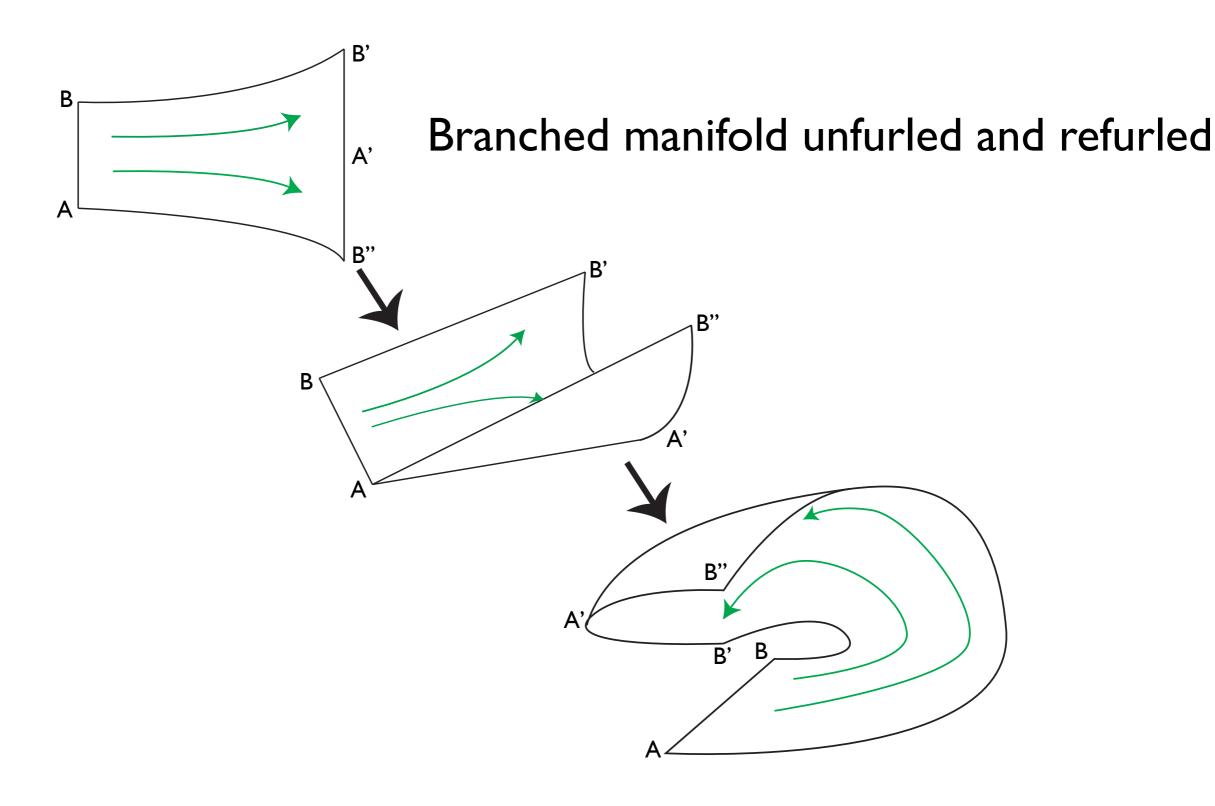
#### Rössler Chaotic Attractor:



#### Branched manifold:



#### Rössler Chaotic Attractor ...



# Rössler stability + instability: Dot spreading demo (ds)

```
Time step = 0.03
Remembered trajectory = 10000
Orient
3000 e
nEns = 100000
IC = (0,-7,0)
radius = .I
I, I, I
```

# Lorenz stability + instability: Dot spreading demo (ds)

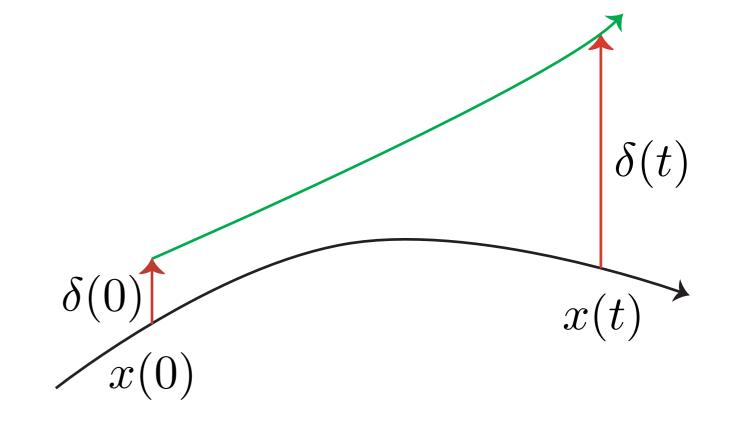
```
Time step = .005
Remembered trajectory = 10000
Orient
1000 e
nEns = 60000
IC = (5,5,5)
radius = .05
I, I, I
```

# Quantifying instability: Growth-of-error model:

$$||\delta(t)|| \sim ||\delta(0)|| e^{\lambda t}$$

Or

$$\lambda \sim t^{-1} \ln \frac{||\delta(t)||}{||\delta(0)||}$$



# Lyapunov Characteristic Exponent (LCE):

$$\lambda = \lim_{\substack{t \to \infty \\ ||\delta(0)|| \to 0}} \frac{1}{t} \log_2 \frac{||\delta(t)||}{||\delta(0)||} \quad \delta(t) \text{ aligns with most unstable direction!}$$

 $\lambda$ : Exponential rate of growth of errors Units: [bits per second]

#### Measurement Resolution: $\epsilon$

Number of scale factors to locate initial state:  $I_0 = -\log_2 \epsilon$ Resolution loss rate (bits per second):  $\lambda$ 

Prediction horizon: 
$$t_{
m unpredict} \sim rac{I_0}{\lambda}$$

# Example:

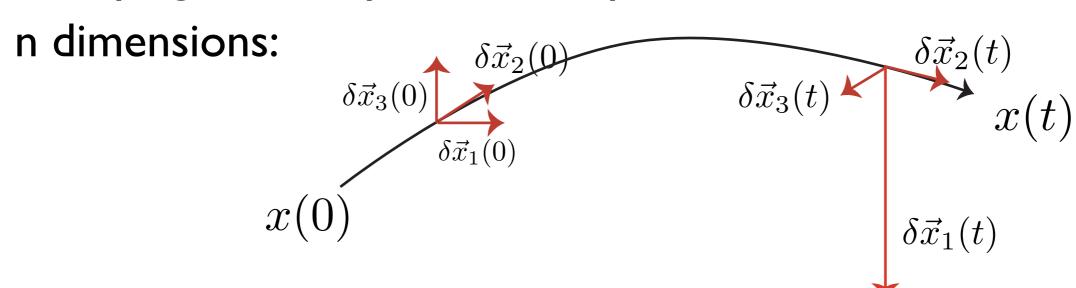
Loss rate: Factor of 2 each second:  $\lambda = 1$ Measurement resolution:  $\epsilon = 10^{-3}$ 

$$I_0 = 10 \text{ bits}$$
  $t_{\text{unpredict}} = 10 \text{ seconds}$ 

Thousand times higher resolution:  $\epsilon = 10^{-6}$ 

$$I_0 = 20 \text{ bits}$$
  $t_{\text{unpredict}} = 20 \text{ seconds}$ 

# Quantifying instability and stability ...



## Lyapunov Characteristic Exponent Spectrum:

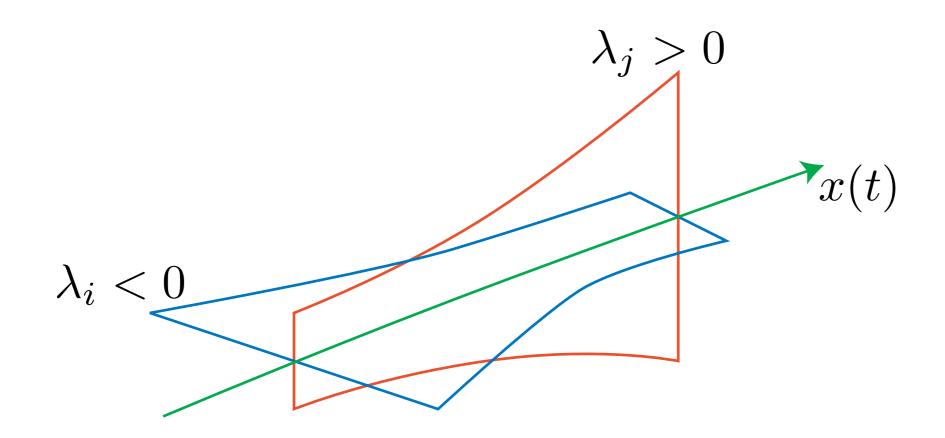
$$\chi = \{\lambda_1, \lambda_2, \dots, \lambda_n\}, \ \lambda_i \ge \lambda_{i+1}$$

$$\lambda_i = \lim_{\substack{t \to \infty \\ ||\delta \vec{x}_i|| \to 0}} \frac{1}{t} \log_2 \frac{||\delta \vec{x}_i(t)||}{||\delta \vec{x}_i(0)||}$$

$$\{\delta \vec{x}_1, \delta \vec{x}_2, \dots, \delta \vec{x}_n\}, \delta \vec{x}_i \cdot \delta \vec{x}_j = 0, i \ne j$$

# Quantifying instability and stability ... LCE Spectrum and Submanifolds:

$$\lambda_i < 0 \iff \text{stable manifold}$$
  
 $\lambda_i > 0 \iff \text{unstable manifold}$ 



LCE Spectrum: Key to characterizing attractors

## Dissipation rate:

## Divergence of vector field:

$$\nabla \cdot \vec{F}(\vec{x}) = \sum_{i=1}^{n} \left. \frac{\partial \vec{F}}{\partial x_i} \right|_{\vec{x}} = \text{Tr}(A(\vec{x}))$$

$$\mathcal{D} = \lim_{T \to \infty} \frac{1}{T} \int_0^T dt \ \nabla \cdot \vec{F}(\vec{x}(t))$$

Theorem: 
$$\mathcal{D} = \sum_{i=1}^{n} \lambda_i$$

$$\chi = \{\lambda_1, \lambda_2, \dots, \lambda_n\}, \ \lambda_i \ge \lambda_{i+1}$$

Dimension of attractor:  $\chi = \{\lambda_1, \lambda_2, \dots, \lambda_n\}, \ \lambda_i \geq \lambda_{i+1}$ 

$$d = j + \frac{\sum_{i=1}^{j} \lambda_i}{|\lambda_{j+1}|}$$

j largest integer such that  $\sum_{i=1}^{j} \lambda_i \geq 0$ 

(Conjectured to be true in most cases.)

Entropy of attractor (jumping ahead a little bit):

$$h_{\mu} = \sum_{\lambda_i > 0} \lambda_i$$

Rate of information production.

#### LCE Attractor Classification:

# An attractor's LCE signature:

$$(\lambda_1, \lambda_2, \dots, \lambda_n), \ \lambda_i \geq \lambda_{i+1}$$

#### Constraints:

I.Attracting:  $\mathcal{D} < 0$ 

(A) 
$$\Rightarrow \sum_{i=1}^{n} \lambda_{i} < 0$$
  $\Rightarrow |\mathcal{D}| > h_{\mu}$   $\Rightarrow \lambda < 0$ , for at least one  $i$   $\Rightarrow \left|\sum_{\lambda_{i} < 0}\right| > \sum_{\lambda_{i} > 0}$ 

2. If not fixed point, flow along trajectory neutrally stable:

$$\lambda_i = 0$$
, for at least one i

3. If chaotic:

$$\lambda_i > 0$$
, for at least one i

## LCE Spectrum Attractor Classification ...

$$(\operatorname{sgn}(\lambda_1),\operatorname{sgn}(\lambda_2),\ldots,\operatorname{sgn}(\lambda_n))$$

Dimension n	LCE Spectrum	Attractor
I	(-)	Fixed Point
2	(-,-)	Fixed Point
2	(0,-)	Limit Cycle
3	(-,-,-)	Fixed Point
3	(0,-,-)	Limit Cycle
3	(0,0,-)	Torus
3	(+,0,-)	Chaotic
4	(0,0,0,-)	3-Torus
4	(+,0,0,-)	Chaotic 2-Torus
4	(+,+,0,-)	Hyperchaos

#### Definition of chaotic attractor:

- (I) Attractor:  $\Lambda \subset \mathcal{X}$ 
  - (a) Invariant set:  $\Lambda = \phi_T(\Lambda)$ .
  - (b) Attracts an open set  $U \subset \mathcal{X}$ :  $\Lambda \subset U$

$$\Lambda = \lim_{T \to \infty} \phi_T(U)$$

- (c) Minimal: no proper subset is also (a) & (b).
- (2) Aperiodic long-term behavior of a deterministic system with exponential amplification.
  - (2') Positive maximum LCE:  $\lambda_{\max}(\Lambda) > 0$ .
  - (2") Positive metric entropy:  $h_{\mu}(\Lambda) > 0$ .

Reading for next lecture:

*NDAC*, Sections 10.5-10.7.

# Thursday Meeting Update

- Invite class to Dept Research
- Overlaps with Thursday class
- 12:10-12:50 PM, 432 Physics
- Then back to 195 Physics