

Last time

- Lyapunov vectors
- singular vectors of tangent / adjoint propagators
- eigenvectors of far past / far future matrices

Future:

- Machine learning + dynamical system ②
- Computational dynamics project ③
- Random dynamical systems ①

Today

- Review of Lyapunov vectors
- Review of QR iteration (linear algebra)
- Computation of LVs.

Kupatsov Parritz 2012 / Trefethen
 Base
 Fix orbit
 $x, \varphi x, \dots, \varphi^k x, \dots$

Far future operator

$$W^+(x) = \lim_{t \rightarrow \infty} \left(\underbrace{(F(0,t))^T}_{\text{tangent propagator}} F(0,t) \right)^{\frac{1}{2t}}$$

$$T_x M \rightarrow T_{\varphi^t x} M$$

$$(F(0,t))^T: T_{\varphi^t x} M \rightarrow T_x M$$

$$F(0,t) = U(0,t) \Sigma(0,t) V(0,t)^T$$

$$\begin{aligned} (F(0,t))^T F(0,t) \\ = V \Sigma U^T U \Sigma V^T = V \Sigma^2 V^T \end{aligned}$$

$$W^+(x) = \lim_{t \rightarrow \infty} \left(\underbrace{(F(0,t))^T F(0,t)}^{1/t} \right)^{\frac{1}{2t}}$$

$$\text{State at time 0} = \lim_{t \rightarrow \infty} V(0,t) \underline{\underline{\Sigma(0,t)^{1/t}}} V(0,t)^T$$

Recall
 $(A^T A, A A^T)$ are SPSP matrices.
 eigenvalues ≥ 0 , eigenvectors are orthogonal

$$V(0,t) := \begin{bmatrix} f_1^+(0,t) & f_2^+(0,t) & \dots & f_d^+(0,t) \end{bmatrix}$$

Existence of
 $\lim_{t \rightarrow \infty} V(0,t)$: Oseledec's Theorem

Logarithms of eigenvalues of
 $W^+(x) := \text{Lyapunov exponents}$

$$\lambda_1 > \lambda_2 > \dots > \lambda_p$$

p : number of distinct LEs

$$p \leq d.$$

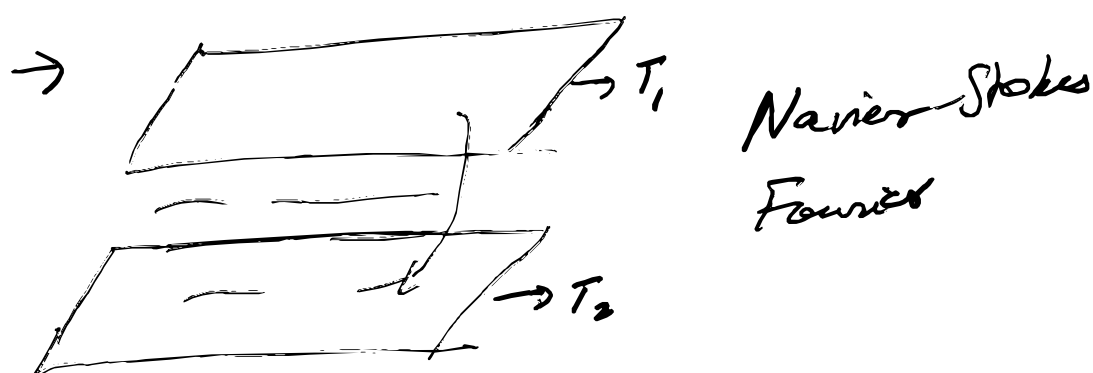
For degenerate LEs, $f_{i(x)}^+$ are not unique

In
 Ergodic systems, LEs are constant functions of x .

$$\lim_{t \rightarrow \infty} f_i^+(0,t) = f_i^+(x) \quad \hookrightarrow \text{at point } x$$

Homework

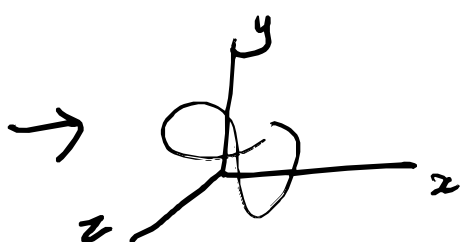
→ Lorenz '63 system : first example of chaos in climate system



→ $F(o, t) : \underset{T_o M}{v(o)} \rightarrow \underset{T_{\phi^t x} M}{v(t)}$

→ $\phi^t x$: solution of Lorenz system at time t

Runge Kutta
odeint (scipy)



$$\frac{d\phi^t(x)}{dt} = g(\phi^t x)$$

→ $v(o) \in T_o M$
 \downarrow
 $v(t) \in T_{\phi^t x} M$

$$\frac{dv(t)}{dt} = dg(\phi^t x) v(t) \quad \leftarrow \text{tangent equation}$$

$$F(o, t) v(o) = v(t)$$

↓
time integrator
for tangent equation

Power iteration

Input:

A

Output: λ_0, v_0 : top eigenvector + value

v_0 not orthogonal to its top eigenvector

for $i = 1, 2, 3, \dots$

$$v_{i+1} = A v_i$$

$$\alpha_i = \|A v_i\|$$

$$v_{i+1} = v_{i+1} / \alpha_i$$

$$v_i \rightarrow v_0 \quad \alpha_i \rightarrow \lambda_0$$

Input:

A_0, A_1, A_2, \dots

Output:

$q_0(\phi^k x), q_0(\phi^{kH} x)$

.....

λ_0 : top LE

$$A_0 = d\phi(x)$$

$$A_1 = d\phi(\phi x) \dots$$

$$v_0 \in T_x M \equiv \text{rand}(d)$$

for $i = 1, 2, 3, \dots$

$$v_{i+1} = A_i v_i \in T_{\phi^i x} M$$

$$\alpha_i = \|A_i v_i\|$$

$$v_{i+1} = v_{i+1} / \|A_i v_i\|$$

$$v_i \rightarrow q_0(\phi^i x)$$

$$\frac{1}{i} \sum_{j=1}^i \log \alpha_j \rightarrow \lambda_0$$

Inputs:

$A_0, A_1, \dots, A_k, \dots$

$$A_0 = (d\phi(\phi^k x))^T$$

$$A_1 = (d\phi(\phi^{k-1} x))^T$$

$$\vdots$$

$$A_k = (d\phi(x))^T$$

$$v_k \in T_x^* M$$

$$\approx p_0(x)$$

QR iteration

GS orthogonalization

Gram-Schmidt

$$\underbrace{\{v_1, \dots, v_n\}}_{\mathbb{R}^d} \rightarrow \{q_1, \dots, q_n\}$$
$$q_i \perp q_j \quad i \neq j$$

$$q_1 \leftarrow v_1 / \|v_1\|$$

$$q_2 = v_2 - \underbrace{(v_2 \cdot q_1)}_{\text{projection}} q_1$$

$$q_2 = q_2 / \|q_2\|$$

$$(q_2 \cdot q_1 = v_2 \cdot q_1 - (v_2 \cdot q_1) = 0)$$

$$q_3 = v_3 - \underbrace{(v_3 \cdot q_1)}_{\text{projection}} q_1 - \underbrace{(v_3 \cdot q_2)}_{\text{projection}} q_2$$

}} modified GS

$$\tilde{q}_3 = v_3 - (v_3 \cdot q_1) q_1$$

$$q_3 = \tilde{q}_3 - (\tilde{q}_3 \cdot q_2) q_2$$

$$\begin{array}{ccccc}
 A & = & Q & R \\
 \downarrow & & \downarrow & \searrow \\
 m \times n & & m \times n & n \times n \\
 & & \downarrow & \searrow \\
 & & \text{orthonormal} & \text{Upper triangular} \\
 & & \text{matrix} & \text{matrix}
 \end{array}$$

Q : form an orthonormal basis for the columns of A

R : off-diagonal elements: orthogonal projections

R : diagonal elements: norms computed by modified GS.

Eigenvalue - Eigenvector of matrix A

$$\begin{array}{c}
 A^k E \cong \text{span}\{e_0, e_1, \dots, e_{k-1}\} \\
 \uparrow \\
 \text{m-dimensional} \\
 \text{subspace} \\
 \text{of } \mathbb{R}^d
 \end{array}$$

Combine power iteration + GS

Input: $A \in \mathbb{R}^{d \times d}$

Q_0 : random orthogonal matrix
 $\in \mathbb{R}^{d \times d}$

R_0 : Id

for $i = 1, 2, \dots$

$$\hat{Q}_{i+1} = A Q_i$$

$$Q_{i+1} R_{i+1} = \hat{Q}_{i+1}$$

$Q_i \rightarrow$ eigenvectors of A

$$\prod_{j=1}^i R_j = R_i R_{i-1} \dots R_0$$

(triangular matrix: eigenvalues are on the diagonal)

$$\begin{aligned}
 A^k &= Q_1 Q_2 \dots Q_k R_k \dots R_1 \\
 &= \tilde{Q}_k \tilde{R}_k
 \end{aligned}$$

$$\begin{array}{l}
 \text{Schur form of } A \\
 = \tilde{Q}_k^* A \tilde{Q}_k
 \end{array}
 \quad \begin{array}{l}
 \text{C triangular, diagonal if} \\
 A \text{ is symmetric}
 \end{array}$$

Rayleigh quotient: v

$$r(v) = v^T A v$$