

Topological Pressure and Dimension

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- The theory of dynamical systems has its origins in the work of Henri Poincaré on the three-body problem of celestial mechanics
- It comprises of a set of tools and methods to study ordinary differential equations and iterated mappings

Example - Quadratic map

$$f : \mathbb{R} \rightarrow \mathbb{R}, f(x) = ax(1-x), a > 0$$

What happens to a point $x \in \mathbb{R}$ if we iterate the function f ?

What is the behaviour of $\{f^n(x)\}_{n \in \mathbb{N}}$?

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One method of studying these problems is by defining new quantities over the systems, these quantities can help us analyse and classify the different dynamics

- Metric Entropy
- Topological Entropy
- Topological Pressure

Loosely speaking, these quantities give us information on how fast points are moving away from each other

- Dimension

In some systems the interesting dynamics is comprised in a complicated subset, this quantity allows us to assign a non integer dimension to these sets

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The Main Example

For the remainder of the presentation we'll consider a particular example

- $\lambda_1, \lambda_2 > 0$ and $\lambda_1 + \lambda_2 < 1$
- $\Delta_1 = [0, \lambda_1], \Delta_2 = [1 - \lambda_2, 1]$
- $\xi_1 = \{\Delta_1, \Delta_2\}$

$$\xi_n = \left\{ \bigcap_{k=0}^{n-1} f^{-k}(\Delta_{i_k}) =: \Delta_{i_0 \dots i_{n-1}} : i_0, \dots, i_{n-1} = 1, 2 \right\}$$

The set of points that remain in the interval $[0, 1]$ forever is

$$K = \bigcap_{n \in \mathbb{N}} \bigcup_{i_1, \dots, i_n} \Delta_{i_1 \dots i_n}$$

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Metric Entropy

Let μ be a probability measure over $[0, 1]$

We define the entropy of each collection ξ_n

$$\xi_1 = \{\Delta_1, \Delta_2\}$$

$$H_\mu(\xi_1) = - \left(\mu(\Delta_1) \log(\mu(\Delta_1)) + \mu(\Delta_2) \log(\mu(\Delta_2)) \right)$$

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Topological Entropy

$$h(f) = \lim_{n \rightarrow \infty} \frac{1}{n} \log(\text{card } \xi_n)$$

- $\text{card } \xi_n = 2^n$
- $h(f) = \lim_{n \rightarrow \infty} \frac{1}{n} \log(2^n) = \log(2)$

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Properties of Entropy

- Entropy is an invariant quantity over the dynamics
- $h_\mu(f^k) = kh_\mu(f)$ and $h(f^k) = kh(f)$ for all $k \in \mathbb{N}$

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Variational Principle

Variational Principle for entropy

$$h(f) = \sup_{\mu} \{h_{\mu}(f)\}$$

$$h(f) \geq \sup_{\mu} \{h_{\mu}(f)\}$$

- It can be shown that $H_{\mu}(\xi_n) \leq \log(\text{card } \xi_n)$ for all $n \in \mathbb{N}$
- $\frac{1}{n} H_{\mu}(\xi_n) \leq \frac{1}{n} \log(\text{card } \xi_n)$
- Taking $n \rightarrow \infty$ we get $h_{\mu}(f) \leq h(f)$

$$h(f) \leq \sup_{\mu} \{h_{\mu}(f)\}$$

- Consider the measure μ that assigns a measure of $\frac{1}{2^n}$ to each element of the collection ξ_n
- $H_{\mu}(\xi_n) = - \sum_{i_1, \dots, i_n} \mu(\Delta_{i_1 \dots i_n}) \log(\mu(\Delta_{i_1 \dots i_n})) = -2^n \frac{1}{2^n} \log(\frac{1}{2^n})$
- $h_{\mu}(f) = \lim_{n \rightarrow \infty} \frac{1}{n} H_{\mu}(\xi_n) = \log(2)$

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Let $\varphi : [0, 1] \rightarrow \mathbb{R}$ be a continuous function

Topological Pressure

$$P(\varphi) = \sup_{\mu} \left\{ h_{\mu}(f) + \int \varphi d\mu \right\}$$

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With this observations we can compute $P(d\varphi) = h_\mu(f) + d \int \varphi d\mu$

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$$h_\mu(f) = \lim_{n \rightarrow \infty} \frac{1}{n} H_\mu(\xi_n) = H_\mu(\xi_1) = -\lambda_1^d \log(\lambda_1^d) - \lambda_2^d \log(\lambda_2^d)$$

$d \int \varphi d\mu$

$$\begin{aligned} d \int \varphi d\mu &= d\mu(\Delta_1) \log(\lambda_1) + d\mu(\Delta_2) \log(\lambda_2) \\ &= d\lambda_1^d \log(\lambda_1) + d\lambda_2^d \log(\lambda_2) \end{aligned}$$

Hausdorff Dimension

Let A be any subset of \mathbb{R}

α -dimensional Hausdorff measure of A

$$m(A, \alpha) = \lim_{\epsilon \rightarrow 0} \inf_{\mathcal{U}} \sum_{U \in \mathcal{U}} (\text{diam } U)^\alpha$$

Where the infimum is taken over all covers \mathcal{U} of A with $|\mathcal{U}| \leq \epsilon$

Hausdorff Dimension

$$\dim_H A = \inf \{ \alpha : m(A, \alpha) = 0 \}$$

- We will compute the Hausdorff dimension of the set K
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- The Topological Entropy and the Metric Entropy are related via the Variational Principle
- The Hausdorff Dimension can be obtained from the Topological Pressure

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- It requires proficiency in a large number of areas such as Topology, Measure Theory, Real Analysis, Differential Geometry
- It uses creative approaches and surprising results to tackle a wide range of complex problems on Dynamics and Differential Equations
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THANK YOU!