

2. Entropy

2. Measure-theoretic entropy

for (X, \mathcal{B}, μ) probability space, the entropy of a partition \mathcal{Z} is given by

$$H(\mathcal{Z}) = H_\mu(\mathcal{Z}) = - \sum_{C \in \mathcal{Z}} \mu(C) \log \mu(C)$$

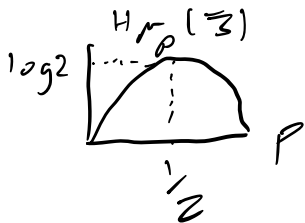
$0 \log 0 := 0$

Remarks: 1) $H(\mathcal{Z}) \geq 0$

2) $H(\mathcal{Z}) = 0 \Leftrightarrow \exists C \in \mathcal{Z}$ with $\mu(C) = 1$

3) If $\mathcal{Z} = \{A_1, \dots, A_k\}$, and $\mu(A_i) = \frac{1}{k}$ for each i , then $H(\mathcal{Z}) = - \sum_{i=1}^k \frac{1}{k} \log \frac{1}{k} = \log k$

$$\mathcal{Z} = \{A_1, A_2\}$$
$$\mu_P(A_1) = P$$



largest possible entropy for partition into k sets

Defn The information function is

$$I(\mathcal{Z})(x) := - \log(\mu(\mathcal{Z}(x)))$$

$$= \sum_{A \in \mathcal{Z}} - \log(\mu(A)) \mathbb{I}_A(x)$$

Note ($I(\mathcal{Z})$ is a measurable fn.) and $I(\mathcal{Z}) \geq 0$

Note that we can write

$$H(\mathcal{Z}) = \int I(\mathcal{Z}) d\mu$$

i.e. $H(\mathcal{Z})$ is average information

For finite partitions $\mathcal{Z}_0, \dots, \mathcal{Z}_{n-1}$

the partition $\bigvee_{i=0}^{n-1} \mathcal{Z}_i$ is determined by:

$$\bigvee_{i=0}^{n-1} \mathcal{Z}_i(x) = \bigwedge_{i=0}^{n-1} \mathcal{Z}_i(x).$$

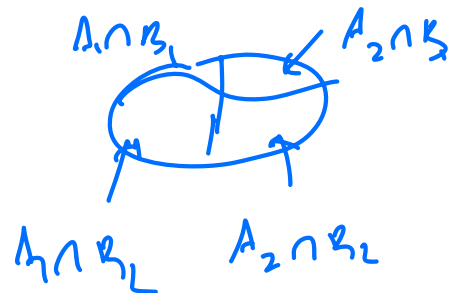
$$\mathcal{Z}_1 = \{A_1, A_2\}$$



$$\mathcal{Z}_2 = \{B_1, B_2\}$$



$$\mathcal{Z}_1 \vee \mathcal{Z}_2$$



We write $\mathcal{Z} \overset{\text{coarser}}{\prec} \mathcal{M} \overset{\text{finer}}{\prec} \mathcal{M}$ if each element of \mathcal{M} is contained in an element of \mathcal{Z} (mod 0)

Good notation because $\#\mathcal{Z} \leq \#\mathcal{M}$

Interpretation

"Guess who?" game. Ask questions to find opponent's identity.

Each yes/no question gives a partition

Denote $I(p)$ as the information received if p is the prob. of a positive answer

Claim: $I(p) = -\log p$ is the natural function that meets criteria to be information:

Want:

- $I(p) > 0$ when $p > 0$
- Choosing a smaller set gives more information
- Information gained by independent results sums

But prob. of positive answer p twice is p^2 . So want $I(p^2) = 2I(p)$

Qn 1: "Is your character female" is
50-50 in the game

Qn 2: "Is your character bald?"
gives a lot of info if yes,
not much if no.

Entropy is the average information
Qn 1 gives more info on
average.

Suppose $X = \{x_1, \dots, x_n\}$ with
respective probabilities (p_1, \dots, p_n) , $\sum p_i = 1$

*p_i is the prob you pick character i to
be your card*
(e.g. Guess who. A good player will
choose in a uniform way
but maybe a player will favor choosing women or
characters with glasses)

Think of a question as a partition

A sequence of questions $\mathcal{E}_1, \dots, \mathcal{E}_n$
is generating if they can

distinguish all results

$$\text{i.e. } \bigcap_{i=1}^n \mathcal{E}_i(x) = \{x\}$$

*partition
indisputably*

By definition, $H(\hat{\bigvee}_{i=1}^n \mathcal{Z}_i) = -\sum p_i \log p_i$

$$\text{Let } h(M) = H(\hat{\bigvee}_{i=1}^n \mathcal{Z}_i)$$

for any generating sequence.

A priori, the average information could depend on the (\mathcal{Z}_i) . It doesn't, by simple computation. So now we can talk about entropy of a measure, not a partition.

Measure entropy

For a measure preserving transformation

$$T: X \rightarrow X$$

$$T^{-1} \mathcal{Z} := \{ T^{-1} A : A \in \mathcal{Z} \}$$

Defn If \mathcal{Z} is a finite partition

of (X, \mathcal{B}, μ) then the entropy of

T w.r.t. \mathcal{Z} is

$$h(\mu, T, \mathcal{Z}) := \lim_{n \rightarrow \infty} \frac{1}{n} H_{\mu}(\bigvee_{i=0}^{n-1} T^i \mathcal{Z})$$

Limit exists because $a_n = H(\bigvee_{i=0}^{n-1} T^{-i} \mathcal{Z})$
 is sub-additive. Limit equals $\inf_n a_n \leq a_1 = H(\mathcal{Z}) < \infty$

Defn The entropy of (X, \mathcal{B}, μ, T) is

$$h(\mu) = \sup_{\mathcal{Z}} h(\mu, T, \mathcal{Z})$$

where sup is over all finite partitions

Want to find \mathcal{Z} s.t. $h(\mu) = h(\mu, T, \mathcal{Z})$

can be checked that if $\mathcal{Z}_n = \bigvee_{i=0}^{n-1} T^{-i} \mathcal{Z}$ or $\bigvee_{i=-n}^n T^{-i} \mathcal{Z}$

satisfies $\text{diam}(\mathcal{Z}_n) \rightarrow 0$ i.e. $\bigcap_{n=1}^{\infty} \mathcal{Z}_n(x) = \{x\} \forall x$

i.e. if the sequence of "questions" \mathcal{Z}_n
 (given by improving \mathcal{Z} using the dynamics)
 sets every outcome, then \mathcal{Z}
 gives the entropy

For a flow $F = (f_t)$ it turns out

$$h(\mu, f_t) = |t| h(\mu, f_1) \quad (\text{Abramov formula})$$

so we let $h(\mu) := h(\mu, f_1)$

2.2 Partitions and symbolic dynamics

$X \subseteq \{0, 1, \dots, k\}^{\mathbb{Z}}$ closed $\sigma(X) = X$

$\sigma: X \rightarrow \{0, 1, \dots, k\}^{\mathbb{Z}}$ shift map

$$\begin{array}{cccccccc}
 x = & \dots & x_{-3} & x_{-2} & x_{-1} & x_0 & x_1 & x_2 & x_3 & \dots \\
 \downarrow \sigma & & \swarrow & \swarrow & \swarrow & \swarrow & \swarrow & & & \\
 y = \sigma(x) = & \dots & x_{-2} & x_{-1} & x_0 & x_1 & x_2 & x_3 & x_4 & \dots
 \end{array}$$

For a shift space X ,

$L = L(X) = \underline{\text{language of } X}$

= the set of all finite words that occur in points of X

$L_n = \text{words in language of length } n$

Natural metric d on X :

$$d(x, y) = \frac{1}{2^k} \text{ where } k \text{ is first index } x \text{ and } y \text{ disagree}$$

if $x_0 \neq y_0, d(x, y) = 1$
 if $x_0 = y_0, d(x, y) = \frac{1}{2}$
 etc..

cylinder sets: If $w = w_1 w_2 \dots w_k$

$${}_p[w] = \{y \in X : y_p \dots y_{p+k-1} = w_1 \dots w_k\}$$

so e.g. ${}_0[i] = \{y \in X : 0^{\text{th}} \text{ entry is } i.\}$

Partitions

$$\text{Let } \mathcal{Z} = \mathcal{Z}_0 = \{ {}_0[0], {}_0[1], \dots, {}_0[k] \}$$

$$\text{Let } \mathcal{Z}_n = \bigvee_{i=-n}^{\hat{}} \sigma^{-i} \mathcal{Z}_0$$

Sets $A \in \mathcal{Z}_n$ are cylinder sets of form

$${}_{-n} [x_{-n} x_{-n+1} \dots x_n]$$

$$\text{Diam}(\mathcal{Z}_n) \leq \frac{1}{2^{n+1}} \rightarrow 0$$

Thus for any σ -inv. prob measure

$$\begin{aligned} \text{Thus } h(\mu, \sigma) &= h(\mu, \sigma, \mathcal{Z}) \\ &= \lim_{n \rightarrow \infty} \frac{1}{n} H_\mu \left(\bigvee_{i=0}^{n-1} \sigma^{-i} \mathcal{Z} \right) \end{aligned}$$

$$h(\mu, \sigma) = \lim_{n \rightarrow \infty} -\frac{1}{n} \sum_{w \in \mathcal{Z}_n} \mu(w) \log \mu(w)$$

where $\mu(w) := \mu({}_0[w]) = \mu({}_p[w]) \forall p \in \mathbb{Z}$
↑ since μ is invariant

Topological entropy of shift space \mathcal{X}

$$h_{\text{top}}(\sigma) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \# \mathcal{Z}_n$$

2.3 Topological entropy, etc

General case (X, f) . X compact metric space
 $f: X \rightarrow X$ cb

Defn A Bowen ball is a set of the form

$$B_n(x, \varepsilon) = \left\{ y \in X : d(f^i x, f^i y) < \varepsilon \quad \forall i \in \{0, \dots, n-1\} \right\}$$

Note: for (Z, σ) , the Bowen balls are cylinder sets

$$\left(\begin{array}{l} \text{Let } d_n(x, y) = \max \{ d(f^i x, f^i y) : i \in \{0, \dots, n-1\} \} \\ \text{Then } d_n \text{ is a metric. Bowen balls are just} \\ d_n\text{-metric balls} \end{array} \right)$$

E is an (n, ε) -spanning set if

$$\bigcup_{x \in E} \overline{B}_n(x, \varepsilon) = X$$

Let $r_n(\varepsilon, X) = \min \{ \#E : E \text{ is an } (n, \varepsilon)\text{-spanning set} \}$

$$\text{Let } \boxed{h_{\text{top}}(f, \varepsilon) := \lim_{n \rightarrow \infty} \frac{1}{n} \log r_n(\varepsilon, X)}$$

$$\text{Let } \boxed{h_{\text{top}}(f) = \lim_{\varepsilon \rightarrow 0} h_{\text{top}}(f, \varepsilon)}$$

Katok entropy formula for $f: X \rightarrow X$

Let $Z \subset X$.

Let $r_n(\epsilon, Z) = \min \{ \#E : E \text{ is } (n, \epsilon)\text{-spanning for } Z \}$

ergodic inv. prob.
↓

i.e. $Z \subseteq \bigcup_{x \in E} \overline{B_n(x, \epsilon)}$

Let $\mu \in M_f^e(X)$. For a fixed $\delta \in (0, 1)$ (e.g. $\delta = \frac{1}{2}$), then

Thm

$$h(\mu, f) = \lim_{\epsilon \rightarrow 0} \overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \log \left(\inf \{ r_n(\epsilon, Z) : Z \text{ has } \mu(Z) \geq 1 - \delta \} \right)$$

Sketch:

Comes from local entropy formula:

for μ -a.e. x $\mu(B_n(x, \epsilon)) \approx e^{-nh(\mu)}$

\therefore Need $\frac{1}{\delta} e^{nh(\mu)}$ such sets to cover a set of measure $1 - \delta$.

or
Brin-Leatley /
Shannon-McMillan-
Breiman in
symbolic
setting

Note: Clear from Katok entropy formula that

$$(*) \sup_{\mu \in M_f^e(X)} h(\mu, f) \leq h_{\text{top}}(f)$$

Recall: Our goal is to study μ which achieve equality in (*)