

(New Chaps) 6 Entropy - History

Entropy

Clausius 1847
macroscopic entropy
in thermodynamics

J.W. Gibbs 1873-78

1916-20, Claude Shannon 1948
Information theory
math theory of communication

Kolmogorov - Sinai ++
1958 1959

metric entropy
measure theoretic entropy

ex upshot: $x \mapsto 2x \pmod{1}$ and $x \mapsto 3x \pmod{1}$
are not measure isomorphic

Def 6.1 Let (X, \mathcal{G}, μ) be a probability space and \mathcal{A} a finite sub-algebra of \mathcal{G} generated by a partition

$$\mathcal{A} = \{A_1, \dots, A_k\} \quad k < \infty$$

The entropy of \mathcal{A} wrt μ is

$$H(\mathcal{A}) = H_\mu(\mathcal{A}) = H(\mathcal{A}) = -\sum_{i=1}^k \mu(A_i) \log \mu(A_i) \geq 0$$

Intuition: H measures the average amount of uncertainty removed (or information gained) by performing a probabilistic experiment

6.2
Lemma The function $\phi(t) = -t \log t$ is strictly concave on $t > 0$.



By Jensen's inequality for $x_i > 0, p_i > 0, \sum p_i = 1$

$$\phi(\sum p_i x_i) \geq \sum p_i \phi(x_i) \quad (*)$$


with equality iff for every i with $p_i > 0$ the x_i 's are equal.

$$H(\mathcal{A}) = \sum_{i=1}^k \phi(\mu(A_i))$$

Proposition 6.3
If $\mathcal{A} = \{A_1, \dots, A_k\}$ then

$H(\mathcal{A}) \leq \log k$
with equality iff $\mu(A_i) = \frac{1}{k}$ all k .

proof: $x_i = \mu(A_i)$ $p_i = \frac{1}{k}$ yields
 $\sum p_i \phi(x_i) \leq \phi(\sum p_i x_i) = \phi(\frac{1}{k})$
" $\frac{1}{k} \sum \phi(x_i) = \frac{1}{k} H(\mathcal{A}) \leq \frac{1}{k} \log k$

Ex: A dice  $X = \{1, \dots, 6\}$
equi-prob

$$H(\mathcal{G}) = \sum_{i=1}^6 \frac{1}{6} \log 6 = \log 6$$

$$\mathcal{A} = \{\text{even, odd}\} = \{(1,3,5), (2,4,6)\}$$

$$H(\mathcal{A}) = \log 2$$

$$\mathcal{B} = \{(1), (2,3), (4,5,6)\}$$

$$H(\mathcal{B}) = \frac{1}{6} \log(6 \cdot 3^2 \cdot 2^3) < \log 3$$

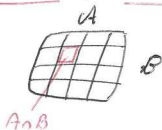
$$= \frac{1}{6} \log 6 + \frac{1}{3} \log 3 + \frac{1}{2} \log 2$$

Def 6.4 Given partitions \mathcal{A}, \mathcal{B}
we define the join of \mathcal{A} and \mathcal{B}

$$\mathcal{A} \vee \mathcal{B} = \{A \cap B : A \in \mathcal{A}, B \in \mathcal{B}\}$$

which is thus a refinement
of both \mathcal{A} and \mathcal{B}

(i.e. every $A \in \mathcal{A}$ is a disjoint
union of elements in $\mathcal{A} \vee \mathcal{B}$)



Def 6.5 We define the conditional
entropy of \mathcal{A} given \mathcal{B} :

$$H(\mathcal{A}|\mathcal{B}) = -\sum_{\substack{A \in \mathcal{A} \\ B \in \mathcal{B}}} P(A \cap B) \log P(A|B)$$

$$= \sum_{A, B} P(B) \Phi(P(A|B)) \geq 0$$

$P(B) P(A|B)$

Prop 6.6 For finite partitions
 \mathcal{A} and \mathcal{B} we have:

$$H(\mathcal{A} \vee \mathcal{B}) = H(\mathcal{A} \vee \mathcal{B}) - H(\mathcal{B})$$

Proof: $H(\mathcal{A}|\mathcal{B}) =$

$$-\sum P(A \cap B) \log \frac{P(A \cap B)}{P(B)} =$$

$$-\sum P(A \cap B) \log P(A \cap B) + \sum P(A \cap B) \log P(B)$$

$$= H(\mathcal{A} \vee \mathcal{B}) + \sum P(B) \log P(B)$$

$$= H(\mathcal{A} \vee \mathcal{B}) - H(\mathcal{B})$$

Coroll 6.7 We have:

$$H(\mathcal{A} \vee \mathcal{B} | \mathcal{B}) = H(\mathcal{A} | \mathcal{B})$$

$$H(\mathcal{B} | \mathcal{B}) = 0$$

since $\mathcal{A} \vee \mathcal{B} \vee \mathcal{B} = \mathcal{A} \vee \mathcal{B}$
 $\mathcal{B} \vee \mathcal{B} = \mathcal{B}$

Example 6.8

Independent partitions:
Suppose $P(A \cap B) = P(A)P(B)$ or
for all A, B with $P(B) > 0$
 $P(A|B) = P(A)$

Then

$$H(\mathcal{A}|\mathcal{B}) = \sum P(B) \Phi(P(A|B))$$

$$= \sum P(B) \Phi(P(A))$$

$$= \sum \Phi(P(A))$$

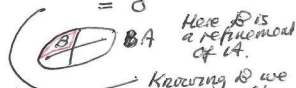
$$= H(\mathcal{A})$$

when \mathcal{A} and \mathcal{B} are indep
knowing \mathcal{B} yields no
information as regards \mathcal{A} .

Refinement. Suppose
 $P(A|B) = 0$ or 1 all A, B

$$H(\mathcal{A}|\mathcal{B}) = -\sum P(B) \Phi(P(A|B))$$

$$= 0$$



knowing \mathcal{B} we
know for sure the result
in \mathcal{A} .

Dice $X = \{1, \dots, 6\}$

\mathcal{B} the power set partition
 $\mathcal{A} = \{(2, 4, 6), (1, 3, 5)\}$
even odd

$$H(\mathcal{B} | \mathcal{A}) = H(\mathcal{B}) - H(\mathcal{A})$$

$$= \log 6 - \log 2$$

$$= \log 3.$$

3 choices left to decide

Prop 6.89

For finite partitions $\mathcal{A}, \mathcal{B}, \mathcal{C}$:

$$H(\mathcal{A} \vee \mathcal{B} | \mathcal{C}) = H(\mathcal{A} | \mathcal{C}) + H(\mathcal{B} | \mathcal{A} \vee \mathcal{C})$$

proof:

$$\begin{aligned} H(\mathcal{A} \vee \mathcal{B} | \mathcal{C}) &= \\ H(\mathcal{A} \vee \mathcal{B} \vee \mathcal{C}) - H(\mathcal{C}) &= \\ H(\mathcal{A} \vee \mathcal{B} \vee \mathcal{C}) - H(\mathcal{A} \vee \mathcal{C}) + \\ H(\mathcal{A} \vee \mathcal{C}) - H(\mathcal{C}) &= \\ H(\mathcal{B} | \mathcal{A} \vee \mathcal{C}) + H(\mathcal{A} | \mathcal{C}) // \end{aligned}$$

mod 0:

Def 6.10. Given finite meas partitions \mathcal{A}, \mathcal{C} we say that $\mathcal{A} \dot{\subseteq} \mathcal{C}$

provided every $A \in \mathcal{A}$ is a finite disjoint union of elements in \mathcal{C} (mod 0)

Thm 6.11 Given finite \mathcal{A}, \mathcal{B} subalgebras of \mathcal{S}_X we have:

- 1) $H(\mathcal{A} | \mathcal{B}) = 0 \iff \mathcal{A} \dot{\subseteq} \mathcal{B}$
- 2) $H(\mathcal{A} | \mathcal{B}) = H(\mathcal{A})$ iff \mathcal{A} and \mathcal{B} are independent:
 $\mu(A \cap C) = \mu(A)\mu(C)$
 $\forall A \in \mathcal{A}, C \in \mathcal{B}$

proof: $H(\mathcal{A} | \mathcal{B}) = 0$ iff for every A, C in the partitions $\mu(A \cap C) \cdot \log \frac{\mu(A \cap C)}{\mu(A)\mu(C)} = 0$ which happens iff $\mu(A \cap C) = 0$ or $\mu(A \cap C) = \mu(A)\mu(C)$ so $C \dot{\subseteq} \mathcal{A}$ (mod 0).

In example 6.8 we showed $\mathcal{A}, \mathcal{B} \text{ indep} \implies H(\mathcal{A} | \mathcal{B}) = H(\mathcal{A})$

Conversely for fixed partition element A :

$$\sum_{C \in \mathcal{B}} \mu(C) \phi(\mu(A|C)) \leq \phi(\mu(A))$$

by Jensen's inequality with equality iff

$$\mu(A|C) = \mu(A) \text{ whenever } \mu(C) > 0.$$

i.e. $\mu(A \cap C) = \mu(A)\mu(C)$ for every partition element A, C whence also for disjoint unions.

later

~~Thm 6.12 $H(\mathcal{B} | \mathcal{A})$ defines a metric on the space of finite subalgebras of \mathcal{S}_X (mod 0)~~

~~proof: $d(\mathcal{A}, \mathcal{B}) \geq 0$ with equality iff $\mathcal{A} \dot{\subseteq} \mathcal{B}$ and $\mathcal{B} \dot{\subseteq} \mathcal{A}$.~~

~~Also $\forall \mathcal{A}, \mathcal{B}, \mathcal{C}$
 $H(\mathcal{A} | \mathcal{C}) \leq H(\mathcal{A} \vee \mathcal{B} | \mathcal{C}) = H(\mathcal{A} | \mathcal{C}) + H(\mathcal{B} | \mathcal{A} \vee \mathcal{C}) \leq H(\mathcal{A} | \mathcal{C}) + H(\mathcal{B} | \mathcal{C})$
 gives the triangle inequality~~

Combined conditional probability

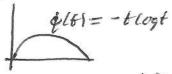
$$P(A \cap B \cap C) = \frac{P(A \cap B \cap C)}{P(B \cap C)} \frac{P(B \cap C)}{P(C)} P(C)$$

$$= P(A|B \cap C) P(B|C) P(C)$$

$\sum_{B \in \mathcal{B}}$

$$\frac{P(A \cap B)}{P(C)} = \sum_{B \in \mathcal{B}} P(A|B \cap C) \frac{P(B|C)}{\sum_B P(B|C)} = 1$$

A.C. fixed



$$\phi(P(A|C)) \geq \sum_B \phi(P(A|B \cap C)) P(B|C)$$

$$P(C) \phi(P(A|C)) \geq \sum_B P(B \cap C) \phi(P(A|B \cap C))$$

Sum over partitions \mathcal{A}, \mathcal{B} :

$$\sum_{A,C} P(C) \phi(P(A|C)) \geq \sum_{A,B,C} P(B \cap C) \phi(P(A|B \cap C))$$

or $H(A|\mathcal{E}) \geq H(A|\mathcal{B} \vee \mathcal{C})$

we have shown:

Prop 6.7

Proposition 6.7.2

Let $\mathcal{A}, \mathcal{B}, \mathcal{C}$ be finite partitions.

Then $H(A|\mathcal{E}) \geq H(A|\mathcal{B} \vee \mathcal{C})$

knowing the outcome of \mathcal{B} and \mathcal{C} leaves less uncertainty for \mathcal{A} than just knowing \mathcal{C} .

The more information we know, a priori, the less we gain by performing an experiment

exs for any \mathcal{E}

$$H(\mathcal{A}) = H(\mathcal{A} | \{\omega, X\}) \geq H(\mathcal{A}|\mathcal{E})$$

$$0 \leq H(\mathcal{A}|\mathcal{X}) \leq H(\mathcal{A}|\mathcal{A}) = 0$$

Corollary 6.8.3 II
Entropy is sub-additive

$$H(\mathcal{A} \vee \mathcal{B}) \leq H(\mathcal{A}) + H(\mathcal{B})$$

$$H(\mathcal{A} \vee \mathcal{B} | \mathcal{E}) \leq H(\mathcal{A}|\mathcal{E}) + H(\mathcal{B}|\mathcal{E})$$

proof: just use

$$H(\mathcal{B}|\mathcal{A} \vee \mathcal{E}) \leq H(\mathcal{B}|\mathcal{E})$$

in Prop 6.9.

Theorem 6.14

$d(\mathcal{A}|\mathcal{E}) = H(\mathcal{A}|\mathcal{E}) + H(\mathcal{E}|\mathcal{A})$ defines a metric on the space of finite sub- σ -alg of \mathcal{X} (mod 0).

$$d(\mathcal{A}|\mathcal{B}) = 0 \iff \mathcal{A} \equiv \mathcal{B}$$

By thm 6.11 $d(\mathcal{A}|\mathcal{E}) = 0$ iff $H(\mathcal{A}|\mathcal{E}) = H(\mathcal{E}|\mathcal{A}) = 0$ iff $\mathcal{A} \subseteq \mathcal{E}$ and $\mathcal{E} \subseteq \mathcal{A}$
 $\Rightarrow \mathcal{A} \equiv \mathcal{E}$

Triangle inequality: ≥ 0
 $H(\mathcal{A} \vee \mathcal{B} | \mathcal{E}) = H(\mathcal{A}|\mathcal{E}) + H(\mathcal{B}|\mathcal{A} \vee \mathcal{E})$
 $= H(\mathcal{B}|\mathcal{E}) + H(\mathcal{A}|\mathcal{B} \vee \mathcal{E})$
 $\leq H(\mathcal{B}|\mathcal{E}) + H(\mathcal{A}|\mathcal{E})$
 $\Rightarrow H(\mathcal{A}|\mathcal{E}) \leq H(\mathcal{A}|\mathcal{B}) + H(\mathcal{B}|\mathcal{E})$
By symmetry
 $H(\mathcal{E}|\mathcal{A}) \leq H(\mathcal{E}|\mathcal{B}) + H(\mathcal{B}|\mathcal{A})$
 $\Rightarrow d(\mathcal{A}, \mathcal{B}) \leq d(\mathcal{A}, \mathcal{C}) + d(\mathcal{C}, \mathcal{B})$

Corollary 6.8.3 I

Let $\mathcal{A}, \mathcal{E}_0, \mathcal{E}_1, \mathcal{E}_2, \dots$ be finite partitions with $\mathcal{E}_0 \subseteq \mathcal{E}_1 \subseteq \mathcal{E}_2 \subseteq \dots$

Then $H(\mathcal{A}|\mathcal{E}_0) \geq H(\mathcal{A}|\mathcal{E}_1) \geq \dots$ (≥ 0)

follows from $\mathcal{E}_{n+1} = \mathcal{E}_n \vee \mathcal{E}_{n+1}$

Entropy of a mean preserving transf.

Let (X, \mathcal{A}, μ, T) be a mean preserving transf of a proba space.

Lemma 6.15 Given finite mean partitions \mathcal{A} and \mathcal{B} :

$$H(T\mathcal{A}) = H(\mathcal{A})$$

$$H(T\mathcal{A} | T\mathcal{B}) = H(\mathcal{A} | \mathcal{B})$$

proof:

When $\mathcal{A} = \{A_1, \dots, A_k\}$ is the \mathcal{A} -partition

then $\mathcal{T}\mathcal{A} = \{T A_1, \dots, T A_k\}$ is the $\mathcal{T}\mathcal{A}$ partition and they have the same probabilities

Same for $\mathcal{A} \vee \mathcal{B}$ for which

$$T(\mathcal{A} \vee \mathcal{B}) = T\mathcal{A} \vee T\mathcal{B} //$$

We may use T to dynamically increase the entropy of a partition

Given $\mathcal{A} = \{A_1, \dots, A_k\}$ we set for $p \geq 1$

$$\mathcal{A}_p = \mathcal{A} \vee T\mathcal{A} \vee \dots \vee T^{(p-1)}\mathcal{A} = \bigvee_{i=0}^{p-1} T^i \mathcal{A}$$

$$\mathcal{A}_0 \equiv \{\emptyset, X\}$$

Def 6.16 We set

$$h_\mu(T, \mathcal{A}) = h(T, \mathcal{A}) = \lim_{p \rightarrow \infty} \frac{1}{p} H(\mathcal{A}_p)$$

It is called the measure theoretic entropy of the map T w.r.t the partition \mathcal{A} and the mean μ .

$$h_\mu(T) = \sup_{\text{finite } \mathcal{A}} h_\mu(T, \mathcal{A}) \in [0, +\infty]$$

is the measure μ -entropy of T .

Theorem 6.17

$$h_\mu(\tau, \mathcal{A}) = \lim_n \frac{1}{n} H_\mu(\mathcal{A}_n) = \inf_{n \geq 0} \frac{1}{n} H_\mu(\mathcal{A}_n) = \liminf_n H_\mu(\mathcal{A} | \bar{T}^n \mathcal{A}_n)$$

(so in particular $\leq H_\mu(\mathcal{A})$)

proof: We have

$$\mathcal{A}_n = \bigvee_0^{n-1} T^{-i} \mathcal{A} = \mathcal{A} \vee T^{-1} \mathcal{A} \vee \dots \vee T^{-(n-1)} \mathcal{A} = \mathcal{A} \vee \bar{T}^n \mathcal{A}_0$$

By corollary 6.13:

$$h_{\mathcal{A}} = H(\mathcal{A}), \quad h_k = H(\mathcal{A} | \bar{T}^k \mathcal{A}_{k-1}), \quad k \geq 1$$

is a non-increasing (positive) sequence so has a limit

$$h_0 \geq h_\infty = \lim_k h_k \geq 0$$

$$\begin{aligned} H(\mathcal{A}_n) &= H(\mathcal{A}_n | \bar{T}^n \mathcal{A}_0) + H(\bar{T}^n \mathcal{A}_0) \\ &= H(\mathcal{A} | \bar{T}^n \mathcal{A}_0) + H(\mathcal{A}_n) \\ &= h_{kn} + H(\mathcal{A}_n) \\ &\dots = \sum_{k=0}^{n-1} h_k = h_0 + \dots + h_1 \quad h_0 = \end{aligned}$$

$$\Rightarrow \frac{1}{n} H(\mathcal{A}_n) \xrightarrow[n \rightarrow \infty]{} h_\infty = h(\tau, \mathcal{A})$$

Entropy inequality and conjugacy invariance.

6.18 $\mathcal{X}_i =$

Theorem: Let $(X_1, \mathcal{B}_1, m_1, T_1)$ and $(X_2, \mathcal{B}_2, m_2, T_2)$ be measure pres. transformations of proba spaces.

a) Suppose that \mathcal{X}_2 is a factor of \mathcal{X}_1 . Then

$$h(\mathcal{X}_2) \leq h(\mathcal{X}_1)$$

b) If \mathcal{X}_1 and \mathcal{X}_2 are isomorphic then

$$h(\mathcal{X}_2) = h(\mathcal{X}_1)$$

proof: Let $\phi: X_1 \rightarrow X_2$ be a meas. pres. isomorphism.

$$\begin{array}{ccc} X_1 & \xrightarrow{T_1} & X_1 \\ \phi \downarrow & & \downarrow \phi \\ X_2 & \xrightarrow{T_2} & X_2 \end{array}$$

More precisely

$$\begin{array}{ccc} \exists A_1 \subset X_1 \text{ full measure } T_1 \text{ inv} \\ \exists A_2 \subset X_2 \text{ full measure } T_2 \text{ inv} \\ A_1 & \xrightarrow{\phi} & A_2 \\ \phi \downarrow & & \downarrow T_2 \\ A_1 & \xrightarrow{T_1} & A_2 \end{array}$$

Pick a finite meas partition α_2 of X_2 .
 $\alpha_2 = \{A_1, \dots, A_r\}$.

Then we may find B_1, \dots, B_r a meas partition of X_1 with $m_1(B_i \cap \phi^{-1}A_j) = 0$, $\alpha_1 = \{B_1, \dots, B_r\}$

Given $i_0, \dots, i_{n-1} \in \{1, \dots, r\}$ we have

$$\phi^{-1} \bigcap_{j=0}^{n-1} T_2^{-j} A_{i_j} = \bigcap_{j=0}^{n-1} T_1^{-j} B_{i_j}$$

so they have the same measure.

Thus
$$H(V_{T_2}^{-n} \alpha) = H(V_{T_1}^{-n} \beta)$$

and taking limits
$$h(T_2, \alpha) = h(T_1, \beta) \leq h(\mathcal{X}_1)$$

whence
$$h(\mathcal{X}_2) = \sup h(T_2, \alpha) \leq h(\mathcal{X}_1)$$

b) \mathcal{X}_1 and \mathcal{X}_2 are factors of each other.

Theorem 6.19

- (1) $h(T, A) \leq h(T, A \vee B) \leq h(T, A) + h(T, B)$
- (2) $h(T, A) \leq h(T, B) + H(\mathcal{A} | \mathcal{B})$
- (3) $h(T, \overline{A}) = h(T, \overline{A}^{-1})$ ($= h(T, T^{-1}A)$ if invertible)
- (4) $h(T, A) = h(T, A_p) \quad \forall p \geq 1$
- (5) When T is invertible:
 $h(T, A) = h(T, \bigvee_{i=-k}^k T^i A) \quad \forall k \geq 0$

Proof:

(1) Use $\bigvee_{i=0}^{n-1} (A \vee B) = (\bigvee_{i=0}^{n-1} A) \vee (\bigvee_{i=0}^{n-1} B)$
 to get
 $H(A_n) \leq H((A \vee B)_n) \leq H(A_n) + H(B_n)$

(2) $H(A_n) \leq H(A_n \vee B_n)$
 $= H(B_n) + H(A_n | B_n)$

$$H(A_n | B_n) \leq \sum_{i=0}^{n-1} H(T^i A | B_n)$$

$$\leq \sum_{i=0}^{n-1} H(T^i A | T^{-i} B_n)$$

$$= n H(A | B)$$

(3) $H(A_n) = H(T^{-1} A_n) = H((T^{-1} A)_n)$

(4) $A_p \vee \dots \vee T^p A_p = A \vee T^{-1} A \vee \dots \vee T^{-p+1} A$
 $= A_{n+p-1}$

$$\frac{1}{n} H(A_{n+p-1}) = \frac{1}{n} H(A_{n+p-1})$$

$$= \frac{n+p-1}{n} \frac{1}{n+p-1} H(A_{n+p-1})$$

$$\rightarrow 1 \times H(T, A)$$

as $n \rightarrow \infty$

(5) $\bigvee_{i=-k}^k T^i A = T^k A_{2k+1}$

Prop 6.20

If \mathcal{A} and \mathcal{B} are finite sub-algebras of \mathcal{B}_X then

$$|h(T, \mathcal{A}) - h(T, \mathcal{B})| \leq$$

$$d(\mathcal{A}, \mathcal{B}) = H(\mathcal{A} | \mathcal{B}) + H(\mathcal{B} | \mathcal{A})$$

In particular $h(T, \cdot)$ is a continuous function on the set of finite sub-algebras (mod 0)

proof Thm 6.19 (2).

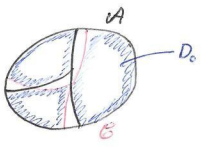
Lemma 6.21

Given $\alpha \in \mathbb{N}$, $\epsilon > 0$ there is $\delta = \delta(k, \epsilon) > 0$ such that:

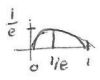
If $\mathcal{F}_A = (A_1, \dots, A_k)$ and $\mathcal{F}_B = (C_1, \dots, C_k)$ are two partitions of card k for which $\mu(A_i \triangle C_i) < \delta \forall i$

Then

$$d(\mathcal{A}, \mathcal{B}) = H(\mathcal{A}|\mathcal{B}) + H(\mathcal{B}|\mathcal{A}) < \epsilon.$$



proof: Recall that $\lim_{t \rightarrow 0^+} \phi(t) = 0$
 $\phi(t) = -t \log t$
 $\phi(1) = 0$



We construct an auxiliary partition: \mathcal{D} consisting of $D_0 = \bigcup_{i=1}^k A_i \cap C_i$ and $A_i \cap C_j, i \neq j$

$$\begin{aligned} \text{Then } 1 &= \sum_i \mu(A_i) = \sum (\mu(A_i \cap C_i) + \mu(A_i \cap C_j)) \\ &\leq \mu(D_0) + k \cdot \delta. \text{ so } \mu(D_0) \geq 1 - k\delta. \end{aligned}$$

Also $\mu(A_i \cap C_j) \leq \mu(A_i \cap C_i^c) < \delta$.

for $i \neq j$

So if $1 - k\delta > 1/2$:

$$\begin{aligned} H(\mathcal{D}) &= \sum_{i \neq j} \phi(\mu(A_i \cap C_j)) + \phi(\mu(D_0)) \\ &\leq k \cdot (k-1) \phi(\delta) + \phi(1 - k\delta) \end{aligned}$$

which goes to zero as $\delta \rightarrow 0$
 So choose δ so that $k\delta < 1/2$ and the RHS $< \epsilon/2$. Then by construction $\mathcal{A} \vee \mathcal{B} = \mathcal{A} \vee \mathcal{D}$ and

$$\begin{aligned} H(\mathcal{A} \vee \mathcal{B}) &= H(\mathcal{A}) + H(\mathcal{B}|\mathcal{A}) \leq \\ &= H(\mathcal{A}) + H(\mathcal{A} \vee \mathcal{D}) \leq H(\mathcal{A}) + H(\mathcal{D}) \leq H(\mathcal{A}) + \epsilon/2 \\ \text{so } H(\mathcal{B}|\mathcal{A}) &< \epsilon/2. \text{ Similarly } H(\mathcal{A}|\mathcal{B}) < \epsilon/2 \end{aligned}$$

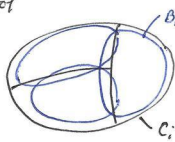
Prop 6.22

- (X, \mathcal{B}_X, μ) probability space.
- $\mathcal{B}_0 \subset \mathcal{B}_X$ a sub-algebra generating $\mathcal{B}_X = \sigma(\mathcal{B}_0)$

Then given any finite sub alg $\mathcal{B} \subset \mathcal{B}_X$ and $\varepsilon > 0$ we may find a finite sub-alg $\mathcal{C} \subset \mathcal{B}_0$ for which

$$d(\mathcal{B}, \mathcal{C}) < \varepsilon.$$

proof



Let $\mathcal{B}_0 = \{C_1, \dots, C_k\}$ be the partition of \mathcal{B}_0 .

Let $\beta > 0$ (to be determined later) and pick B_i

$B_i \in \mathcal{B}_0$ such that $\mu(B_i \Delta C_i) \leq \beta$.

The collection $\{B_1, \dots, B_k\}$ is an approximation of \mathcal{B}_0 but not a partition. Define

$$A_1 = B_1$$

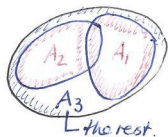
$$A_2 = B_2 \cap B_1^c$$

$$A_3 = B_3 \cap B_2^c \cap B_1^c$$

$$\vdots$$

$$A_{k-1} = B_{k-1} \cap B_{k-2}^c \cap \dots \cap B_1^c$$

finally $A_k = B_k \cap B_{k-1}^c \cap B_{k-2}^c \cap \dots \cap B_1^c$



↳ the rest.

For notational simplicity write $|u| = \mu(u)$.

$$|A_j \Delta C_j| = |B_j \Delta C_j| \leq \beta$$

For $j < k$ we have

$$|A_j \cap C_j^c| \leq |B_j \cap C_j^c|$$

$$\begin{aligned} C_j \cap A_j^c &= C_j \cap (B_j^c \cup \bigcup_{i \neq j} B_i) \\ &= \underline{C_j} \cap (B_j^c \cap \bigcup_{i \neq j} B_i \cap \underline{C_j^c}) \\ &\subset (C_j \cap B_j^c) \cup \bigcup_{i \neq j} (B_i \cap C_j^c) \end{aligned}$$

$$\text{So } |C_j \Delta A_j^c| \leq j \cdot \beta$$

For $j = k$ we have

$$\begin{aligned} C_k \cap A_k^c &= C_k \cap \bigcup_{i < k} B_i = \\ &= \underline{C_k} \cap \bigcup_{i < k} (B_i \cap \underline{C_k^c}) \text{ so} \\ |C_k \cap A_k^c| &\leq \sum_{i < k} |B_i \cap C_k^c| \end{aligned}$$

$$\begin{aligned} A_k \cap C_k^c &= (\bigcap_{i < k} B_i^c) \cap (\bigcup_{j < k} B_j) \\ &\subset \bigcup_{j < k} (C_j \cap B_j^c) \text{ and} \end{aligned}$$

$$|A_k \cap C_k^c| \leq \sum_{j < k} |C_j \cap B_j^c|$$

which shows:

$$|A_k \Delta C_k| \leq (k-1)\beta$$

Choosing

$$\beta = \frac{\varepsilon}{k-1} \delta(k, \varepsilon)$$

with $\delta(k, \varepsilon)$ as in Lemma 6.21 we obtain the wanted bound. //

Thm 6.23 [Kolmogorov-Sinai]
58-59

Let $(X, \mathcal{B}_X, \mu, T)$ be a meas pres transf of a probability space.

1. Suppose that $\mathcal{A} \subset \mathcal{B}_X$ is a finite subalgebra with

$$\bigvee_{i=0}^{\infty} T^i \mathcal{A} \stackrel{\text{a.e.}}{=} \mathcal{B}_X$$

Then $h_\mu(\sigma) = h_\mu(T, \mathcal{A})$

2. Let T be invertible and suppose $\mathcal{A} \subset \mathcal{B}_X$ is finite and

$$\bigvee_{i=-\infty}^{\infty} T^i \mathcal{A} \stackrel{\text{a.e.}}{=} \mathcal{B}_X$$

Then $h_\mu(\sigma) = h_\mu(T, \mathcal{A})$

1. proof: We let $A_n = \bigvee_{i=0}^{n-1} T^i \mathcal{A}$

The hypothesis is that

$$\mathcal{B}_0 = \bigcup_{n \geq 1} A_n \text{ generates } \mathcal{B}_X = \sigma(\mathcal{B}_0).$$

Let $\epsilon > 0$ and $\xi_\epsilon = \{C_1, \dots, C_k\}$

be a finite partition. We may then by Thm 6.22 find

$$\mathcal{A} \ni A_1, \dots, A_k \in \mathcal{B}_0$$

such that $\eta = \{A_1, \dots, A_k\}$ satisfies

$$H(\mathcal{B} | \eta) < \epsilon$$

Since k is finite we have

$$\eta \subset \mathcal{A}_p \text{ for some } p \geq 1$$

$$\begin{aligned} \text{Then } h(T, \mathcal{B}) &\leq h(T, \mathcal{A}_p) + H(\xi_\epsilon | \mathcal{A}_p) \\ &= h(T, \mathcal{A}) + H(\mathcal{B} | \mathcal{A}_p) \\ &\leq h(T, \mathcal{A}) + H(\mathcal{B} | \eta) \\ &= h(T, \mathcal{A}) + \epsilon \end{aligned}$$

so as $\epsilon > 0$ was arbitrary

$$h_\mu(\sigma) = \sup_{\mathcal{B} \text{ finite}} h_\mu(T, \mathcal{B}) \leq h_\mu(T, \mathcal{A}) \leq h_\mu(\sigma).$$

2. In this case we let $A_n = \bigvee_{i=-n}^n T^i \mathcal{A}$ and

note that again

$$\mathcal{B}_0 = \bigcup_{n \geq 0} A_n$$

generates \mathcal{B}_X .

The remaining part is the same.

Entropy of a shift

$$\Sigma_d^+ = \{ \bar{s} = (s_i)_{i \geq 0} : 1 \leq s_i \leq d \}$$

$$T = \sigma^+ : \Sigma_d^+ \rightarrow \Sigma_d^+ \quad \text{shift}$$

$\vec{p} = (p_1, \dots, p_d)$ probability vector

Define the measure on cylinders:

$$\mu_{\vec{p}}(I_{b_0 \dots b_{k-1}}) = p_{b_0} \dots p_{b_{k-1}}$$

$$\{ \bar{s} \in \Sigma_d^+ : s_0 = b_0, \dots, s_{k-1} = b_{k-1} \}$$

and let $\mu_{\vec{p}}$ be the Kolmogorov extended measure on Σ_d^+ , invariant and ergodic under σ^+ .

Let $\mathcal{A} = \{ I_1, \dots, I_d \}$ be the initial partition into elementary cylinders.

Then

$$\mathcal{A}_k = \sigma^+ \mathcal{A} \dots \sigma^{-(k-1)} \mathcal{A} = \{ I_{b_0 \dots b_{k-1}} \}$$

is the collection of k -cylinders in Σ_d^+ . They form a base for the topology, whence generate the Borel σ -algebra.

$$\mathcal{B} = \sigma(\cup_{k \geq 1} \mathcal{A}_k)$$

Since $\mu(I_{b_0 \dots b_{k-1}}) = \mu(I_{b_0}) \dots \mu(I_{b_{k-1}})$ we see that $\mathcal{A}, \sigma^+ \mathcal{A}, \dots, \sigma^{-(k-1)} \mathcal{A}_{k-1}$ are independent sub-algebras. Thus,

$$H(\sigma^+ \dots \sigma^{-(k-1)} \mathcal{A}) = H(\mathcal{A}) + \dots + H(\sigma^{-(k-1)} \mathcal{A})$$

$$= k H(\mathcal{A})$$

$$= k \sum_{i=1}^d p_i \log \frac{1}{p_i} \quad \text{and}$$


$$K-S \text{ thm} \Rightarrow h_{\mu_{\vec{p}}}(T) = \sum_{i=1}^d p_i \log \frac{1}{p_i}$$

$$T_2 x = 2x \pmod{1}$$

Leb meas: λ

$$(S^1, \lambda, T_2) \simeq (\Sigma_2^+, \mu_{(\frac{1}{2}, \frac{1}{2})}, \sigma^+)$$

isom




$$h_{\lambda}(T_2) = \frac{1}{2} \log 2 + \frac{1}{2} \log 2$$

$$= \log 2$$

$$T_3 x = 3x \pmod{1}, \text{ Leb } \lambda$$

$$(S^1, \lambda, T_3) \simeq (\Sigma_3^+, \mu_{(\frac{1}{3}, \frac{1}{3}, \frac{1}{3})}, \sigma^+)$$



$$h_{\lambda}(T_3) = \log 3$$

Conclusion

(S^1, λ, T_2) and (S^1, λ, T_3) cannot be isomorphic.